Remarks

Reconsideration of this Application is respectfully requested.

Claims 38-42, 45-49 and 51-57 are pending in the application, with claims 38 and 51 being the independent claims. Claims 1-37, 43, 44 and 50 are cancelled. Claims 38, 47, 48, 49 and 51 have been amended. Support for the amendments to claims 38 and 51 claims may be found throughout the specification. Support for the amendment to claims 47-49 may be found throughout the specification, e.g., in paragraph [0072] on page 23.

Based on the following remarks, Applicants respectfully request that the Examiner reconsider all outstanding rejections and that they be withdrawn.

Examiner Interview

On June 6, 2007, the Examiner initiated a telephonic interview with Applicants representatives. The Examiner suggested additional amendments to the claims to place the claims in condition for allowance. Applicants thank the Examiner for suggesting the claim amendments and have amended the claims in accordance with the Examiner's suggestions.

The Examiner further requested that Applicants submit copies of Declarations Under 37 C.F.R. § 1.132 by En Li, Ph.D., an inventor, and Kenneth D. Bloch, M.D. that were filed in parent Appl. No. 09/720,086 in support of Applicants' priority claim. Copies of these documents are submitted herewith.

Accordingly, Applicants respectfully request that the Examiner reconsider and withdraw all outstanding objections and rejections.

Conclusion

All of the stated grounds of objection and rejection have been properly traversed, accommodated, or rendered moot. Applicants therefore respectfully request that the Examiner reconsider all presently outstanding objections and rejections and that they be withdrawn. Applicants believe that a full and complete reply has been made to the outstanding Office Action and, as such, the present application is in condition for allowance. If the Examiner believes, for any reason, that personal communication will expedite prosecution of this application, the Examiner is invited to telephone the undersigned directly at (202) 772-8658.

Prompt and favorable consideration of this Reply is respectfully requested.

Respectfully submitted,

STERNE, KESSLER, GOLDSTEIN & FOX P.L.L.C.

Daniel J. Nevrivy
Agent for Applicants

Registration No. 59,118

Date:

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THE UNITED STATES PATENT AND TRADEMARK OFFICE

In re application of:

Li et al.

Appl. No. 09/720,086

102(e): July 23, 2001

De Novo DNA Cytosine Methyltransferase Genes,

Polypeptides and Uses Thereof

Confirmation No.: 6968

Art Unit:

1642

Examiner: Harris, A. M.

Atty. Docket: 0609.4560002/KRM/DJN

Declaration Under 37 C.F.R. § 1.132 of En Li, Ph.D.

Commissioner for Patents P.O. Box 1450 Alexandria, VA 22313-1450

Sir:

- L the undersigned, En Li, Ph.D., residing at 45 Hinckley Road, Newton, Massachusetts, 02168, declare and state as follows:
 - 1. I am a co-inventor of the above-captioned patent application.
- I am currently employed by Novartis Institutes for Biomedical Research as Vice President & Global Head, Animal Models of Disease and Epigenetics Program. Prior to my current employment, I was an Associate Professor of Medicine at Harvard Medical School and directed a laboratory in the Cardiovascular Research Center at the Massachusetts General Hospital from January 1993 to April 2003, where I conducted and supervised research in the field of mouse genetics and developmental biology.
 - 3. A current curriculum vitae is appended hereto as EXHIBIT A.
- I have reviewed the above-captioned patent application and the Office Action dated June 6, 2005. I have also reviewed the sequence listing as filed and the sequence listing as amended on July 23, 2001. I have also reviewed the claims of the captioned patent application.
- I have been informed that the Examiner has not granted priority to the earlier filed patent applications because there is insufficient proof that the coding regions of currently amended SEQ ID NOS:1 and 2 are the same as those listed in the priority documents, viz., the mouse Dnmt3a and Dnmt3b cDNA clones encoding the coding regions of SEQ ID NOS:1 and 2, respectively, that were deposited with the American Type Culture Collection (ATCC), 10801 University Boulevard, Manassas, Virginia 20110-2209, USA. Sequences harboring the coding regions of SEQ ID NOS: 1 and 2, respectively, were

deposited with the ATCC on June 16, 1998, and assigned ATCC Deposit Nos. 209933 and 209934, respectively. The deposit date of June 16, 1998 was prior to the filing date of the first provisional application, App. No. 60/090,906, filed June 25, 1998, the benefit of which is claimed. The '906 application includes the sequence information and references the deposits of the sequenced material on page 15, lines 26, through page 16, line 2, of the specification.

- 6. In November 2004, Applicants had samples withdrawn of the mouse Dnmt3a and Dnmt3b cDNA clones contained within ATCC Deposit Nos. 209933 and 209934, respectively. At the Applicants request, Kenneth D. Bloch, M.D., a faculty member in the Cardiovascular Research Center at the Massachusetts General Hospital and an experienced DNA sequencer, sequenced nucleotides that spanned the coding regions of mouse Dnmt3a and Dnmt3b in the deposited cDNA. A nucleotide alignment that spans the coding regions of sequenced mouse Dnmt3a cDNA clone contained in ATCC Deposit No. 209933 and currently amended SEQ ID NO:1 is shown in EXHIBIT B. A nucleotide alignment that spans the coding regions of sequenced mouse Dnmt3b cDNA clone contained in ATCC Deposit No. 209934 and currently amended SEQ ID NO:2 is shown in EXHIBIT C.
- 7. The amendment to the sequence listing, which was filed on July 23, 2001, corrected six nucleotides in the coding sequence of SEQ ID NO:1 (see the bolded nucleotides at positions 516, 843, 1036, 1110, 1116 and 1726 in EXHIBIT B) and two nucleotides in the coding sequence of SEQ ID NO:2 (see the bolded nucleotides at positions 918 and 920 in EXHIBIT C).
- 8. The deposited clones recited in ¶5 and 6, above (i.e., ATCC Deposit Nos. 209933 and 209934) are the same as the deposited clones recited in the above-captioned application. The coding sequence of ATCC Deposit No. 209933 is currently believed to be the same as the coding sequence of currently amended SEQ ID NO:1. The coding sequence of ATCC Deposit No. 209934 is currently believed to be the same as the coding sequence of currently amended SEQ ID NO:2.
- 9. It is well known that sequencing errors are a common problem in Molecular Biology. See, e.g., Peter Richterich, Estimation of Errors in 'Raw' DNA Sequences: A Validation Study, 8 Genome Research 251-59 (1998) (EXHIBIT D). I believe that one skilled in the art would have sequenced the deposited material and recognized the sequencing errors in the coding region. I believe that the correct mouse Dnmt3a and Dnmt3b coding sequences are inherent to the ATCC deposited clones, ATCC Deposit Nos. 209933 and 209934, respectively, which were deposited prior to the filing of App. No. 60/090,906, filed June 25, 1998, the benefit of which is claimed.
- 10. Accordingly, based on the above, I believe that Applicants are entitled to the June 25, 1998 filing date for the coding sequences of mouse Dnmt3a and Dnmt3b contained within ATCC Deposit Nos. 209933 and 209934, respectively.

- 3 -

Li et al. Appl. No. 09/720,086

11. I hereby declare that all statements made herein of my own knowledge are true and that all statements made on information and belief are believed to be true; and further that these statements were made with the knowledge that willful false statements and the like so made are punishable by fine or imprisonment, or both, under Section 1001 of Title 18 of the United States Code and that such willful false statements may jeopardize the validity of the present patent application or any patent issued thereon.

Respectfully submitted,

En Li, Ph.D.

Date: 11/7/05

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EXHIBIT A

CURRICULUM VITAE

En Li, Ph.D.

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Animal Models of Disease
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Education:

1984	B.Sc. in Biochemistry. Peking University, Beijing, China
1992	Ph.D. Massachusetts Institute of Technology (Biology)
	(Advisor: Prof. Rudolf Jaenisch)

Professional Experience:

1993-1996	Principal Investigator, Cardiovascular Research Center, Massachusetts General
	Hospital
	Instructor, Department of Medicine, Harvard Medical School
1996-2000	Assistant Professor, Department of Medicine, Harvard Medical School
2000-2003	Associate Professor, Department of Medicine, Harvard Medical School
2001-2003	Guest Professor, Beijing University, Health Science Center
2003-	VP & Global Head, Models of Disease Center, Epigenetics Program
	Novartis Institute for Biomedical Research

Review and editorial board

1999-	External Grant Reviewer, Human Frontier Science Program
1999-	External Grant Reviewer, NIH, NICHD
1999-	Member of the Advisory Board. Journal of Biochemistry
2000-	External Grant Reviewer, NIH, NIA
2000-	External Grant Reviewer, NSF
2000-	Mail Reviewer, Wellcome Trust
2003-	Member of the Advisory Board. China Science Reports
2004	External Grant Reviewer, Chinese Natural Science Foundation.
1993-	Ad Hoc Reviewer for the following journals Nature, Science, Cell, Nat Genet, Genes Dev, Trends Genet, Development, Mol Cell Biol, PNAS, Human Mol. Genet., Dev. Biol., Mech. Dev., J. Cell Biol., Dev. Dyn., Gene, Genomics, Nucl Acid Res, Mammalian Genome, etc.

Professional Societies

1990-	American Association for the Advancement of Science, Member
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Ray Wu Society, Member 1999-**Invited Presentations (Since 2003-)** Invited speaker, Gordon Research Conference - Cancer Genetics and Epigenetics 2003 2003 Session chair, Keystone Symposium - Chromatin (Big Sky, Montana) 2003 Invited speaker, Annual HUGO meeting at Cancun, Mexico Invited speaker, Gordon Research Conference - Epigenetics 2003 Invited speaker, the 2nd Annual CDB symposium, Kobe, Japan Invited speaker, 2nd Weiseenburg Symposium on DNA methylation – an important genetics 2004 2004 signals. Weissenburg, Germany Session Chair, on genomic imprinting. 10th SCBA International Symposiums, Beijing, China 2004 Invited speaker, Genomic imprinting workshop in Montpellier, France 2004 2005 Vice Chair, Gordon Research Conference 'Cancer Genetics and Epigenetics'

DNA Methylation Society, Member

1994-

Publication:

- 1. Zijlstra M, Li E, Sajjadi F, Subramani S, Jaenisch R. Germ-line transmission of a disrupted b2-microglobulin gene produced by homologous recombination in embryonic stem cells. Nature 1989; 342: 435-8.
- 2. Lee K, Li E, Huber J, Landis S, Sharpe A, Chao MV, Jaenisch R. Targeted mutation of the p75 low affinity NGF receptor gene leads to deficits in the peripheral sensory nervous system. Cell 1992; 69: 737-49.
- 3. Li E, Bestor TH, Jaenisch R. Targeted mutation of the DNA methyltransferase gene results in embryonic lethality. Cell 1992; 69: 915-26.
- 4. Li E, Sucov HM, Lee K, Evans RM, Jaenisch R. Normal development and growth of mice carrying a targeted disruption of the all retinoic acid receptor gene. Proc Natl Acad Sci USA 1993; 90: 1590-4.
- 5. Li E, Beard C, Jaenisch R. Role for DNA methylation in genomic imprinting. Nature 1993; 366: 362-5.
- 6. Jüttermann R, Li E, Jaenisch R. The toxicity of 5-aza-2'-deoxycytidine to mammalian cells is mediated primarily by DNA methyltransferase rather than DNA demethylation. Proc Natl Acad Sci USA 1994; 91, 11797-11801.
- 7. Laird PW, Jackson-Grusby L, Fazeli A, Dickinson W, Jung E, Li E, Weinberg RA, Jaenisch R. Suppression of intestinal neoplasia by DNA hypomethylation. Cell 1995; 81:197-205.
- 8. Shinoda K, Lei H, Yoshii H, Nomura M, Nagano M, Shiba H, Sasaki H, Osawa Y, Ninomiya Y, Niwa O, Morohashi K, Li E. Developmental defects of the ventromedial hypothalamic nucleus and pituitary gonadotroph in the Ftz-F1 disrupted mice. Dev Dyn 1995; 204: 22-9.
- 9. Beard C, Li E, Jaenisch J. Loss of methylation activates Xist in somatic but not in embryonic cells. Genes Dev 1995; 9: 2325-2334.
- 10. Nan X, Tate P, Li E, Bird A. DNA methylation specifies chromosomal localization of MeCP2. Mole Cell Biol 1996; 16: 414-421.
- 11. Zhang W, Zimmer G, Chen J, Ladd D, Li E, Alt FW, Wiederrecht G, Cryan J, O'Neill EA, Seidman CE, Abbas AK, Seidman JG. T cell responses in calcineurin Aα-deficient mice. J Exp Med 1996; 183: 413-420.
- 12. Tucker KL, Beard C, Dausman J, Jackson-Grusby L, Laird PW, Lei H, Li E, Jaenisch R. Germ-line passage is required for establishment of methylation and expression patterns of imprinted but not of nonimprinted genes. Genes Dev 1996; 10: 1008-1020.
- 13. Trasler JM, Trasler DG, Bestor TH, Li E, Ghibu F. Nuclear localization and expression of DNA methyltransferase in embryos is temporally correlated with postimplantation increase in DNA methylation. Dev Dyn 1996; 206: 239-247.

- 14. Harbers K, Müller U, Grams A, Li E, Jaenisch R, Franz T. Provirus integration into a gene encoding a ubiquitin-conjugating enzyme results in a placental defect and embryonic lethality. Proc Natl Acad Sci USA 1996; 93: 12412-7.
- 15. Nakamuta M, Chang B, Zsigmond E, Kobayashi K, Lei H, Ishida BY, Oka K, Li E, Chan L. Complete phenotypic characterization of apobec-1 knockout mice with a wild-type genetic background and a human apolipoprotein B transgenic background, and restoration of apolipoprotein B mRNA editing by somatic gene transfer of apobec-1. J Biol Chem 1996; 271: 25981-8.
- 16. Lei H, Oh SP, Okano M, Juttermann R, Goss KA, Jaenisch R, Li E. De novo DNA cytosine methyltransferase activities in mouse embryonic stem cells. Development 1996; 122: 3195 3205.
- 17. Chae T, Kwon YT, Bronson R, Dikkes P, Li E, Tsai L-H. Mice lacking p35, a neuronal specific activator of Cdk5, display cortical lamination defects, seizures, and adult lethality. Neuron 1997; 18: 29-42.
- 18. Simon AM, Goodneough DA, Li E, Paul DL. Female infertility in mice lacking connecxin 37. Nature 1997; 385: 525-9.
- 19. Oh SP, Li E. The signaling pathway mediated by type IIB activin receptor controls axial patterning and lateral asymmetry in the mouse. Genes Dev 1997; 11: 1812-1826.
- 20. Gong X, Li E, Klier G, Huang Q, Wu Y, Lei H, Kumar NM, Horwitz J, Gilula NB. Disruption of a3 connexin gene leads to protein degradation and cataractogenesis in mice. Cell 1997; 91: 833-843.
- 21. Bonventre JV, Huang Z, Taheri MR, O'Leary E, Li E, Moskowitz MA, Sapirstein, A. Cytosolic phospholipase A2 deficient mice have abnormal fertility and are protected against focal ischemic injury to the brain. Nature 1997; 390: 622-5.
- 22. Pradhan S, Talbot D, Sha M, Benner J, Hornstra L, Li E, Jaenisch R, Roberts RJ. Baculovirus-mediated expression and characterization of the full-length murine DNA methyltransferase. Nucl Acids Res 1997; 25: 4666-4673.
- 23. Wang S, Miura M, Jung Y-K, Zhu H, Li E, Yuan J. Murine caspase-11, an ICE-interacting protease is essential for the activation of ICE. Cell 1998; 92: 501-9.
- 24. Gu Z, Nomura M, Simpson BB, Lei H, van den Eijnden-van Raaij AJM, Donahoe PK, Li E. The type I activin receptor ActR-IB is required for egg cylinder organization and gastrulation in the Mouse. Genes Dev 1998; 12: 844-857.
- 25. Okano M, Xie S, Li E. Dnmt2 is not required for de novo and maintenance methylation of viral DNA in embryonic stem cells. Nucl Acids Res 1998; 26: 2536-2540.
- 26. Bergeron L, Perez GI, MacDonald G, Shi L, Sun Y, Jurisicova A, Varmuza S, Latham K.E Flaws JA, Salter JC, Hara H, Moskowitz MA, Li E, Greenberg A, Tilly JL, Yuan J. Defects in regulation of apoptosis in caspase-2-deficient mice. Genes Dev 1998; 12: 1304-1314.

- 27. Yao TP, Oh SP, Fuchs M, Zhou ND, Ch'ng LE, Newsome D, Bronson RT, Li E, Livingston DM, Eckner R. Gene dosage-dependent embryonic development and proliferation defects in mice lacking the transcriptional integrator p300. Cell 1998; 93: 361-372.
- 28. Chung UI, Lanske B, Lee K, Li E, Kronenberg, H. The parathyroid hormone/parathyroid hormone-related peptide receptor coordinates endochondral bone development by directly controlling chondrocyte differentiation. Proc Natl Acad Sci USA 1998; 95: 13030-5.
- 29. Nomura M, Li E. Roles for Smad2 in mesoderm formation, left-right patterning, and craniofacial development in mice. Nature 1998; 393: 786-790.
- 30. Okano M, Xie S, Li E. Cloning and characterization of a family of novel mammalian DNA (cytosine-5) methyltransferases. Nat Genet 1998; 19: 219-220.
- 31. Morohashi K, Tsuboi-Asai H, Matsushita S, Suda M, Nakashima M, Sasano H, Hataba Y, Li CL, Fukata J, Irie J, Watanabe T, Nagura H, Li E. Structural and functional abnormalities in the spleen of an mFtz-F1 gene-disrupted mouse. Blood 1999; 93: 1586-1594.
- 32. Gu Z, Reynolds EM, Song J, Lei H, Feijen A, Yu L, He W, MacLaughlin DT, van den Eijnden-van Raaij J, Donahoe PK, Li, E. The type I serine/threonine kinase receptor ActRIA (ALK2) is required for gastrulation of the mouse embryo. Development 1999; 126: 2551-2561.
- 33. Verheijen MH, Karperien M, Chung U, van Wijuen M, Heystek H, Hendriks JA, Veltmaat JM, Lanske B, Li E, Lowik CW, de Laat SW, Kronenberg HM, Defize LH. Parathyroid hormone-related peptide (PTHrP) induces parietal endoderm formation exclusively via the type I PTH/PTHrP receptor. Mech Dev 1999; 81: 151-161.
- 34. Song J, Oh SP, Schrewe H, Nomura M, Lei H, Okano M, Gridley T, Li E. The type II activin receptors are essential for egg cylinder growth, gastrulation and rostral head development in mice. Dev Biol 1999; 213: 157-169.
- 35. Xie S, Wang Z, Okano M, Nogami M, Li Y, He W, Okumura K, Li E. Cloning, expression, and chromosome locations of the human DNMT3 gene family. Gene 1999; 236: 87-95.
- 36. Donohoe MA, Zhang X-L, McGinnis L, Biggers J, Li E, Shi Y. Targeted disruption of mouse Yin Yang 1 transcription factor results in peri-implantation lethality. Mol Cell Biol 1999; 19: 7237-7244.
- 37. Okano M, Bell DW, Haber DA, Li E. DNA methyltransferases Dnmt3a and Dnmt3b are essential for de novo methylation and mammalian development. Cell 1999; 99: 247-257.
- 38. Okano M, Takebayashi S, Okumura K, Li E. Assignment of cytosine-5 DNA methyltransferases Dnmt3a and Dnmt3b to mouse chromosome bands 12A2-A3 and 2H1 by in situ hybridization. Cytogenet Cell Genet 1999; 86: 333-334.
- 39. Li YP, Chen W, Liang Y, Li E, Stashenko P. OC-116Kda-deficient mice exibit severe osteopetrosis due to loss of osteoclast-mediated extracellular acidification. Nat Genet 1999; 23: 452-456.

- 40. Nakagawa T, Zhu H, Morishima N, Li E, Xu J, Yankner BA, Yuan J. Caspase-12 mediates endoplasmic-reticulum-specific apoptosis and cytotoxicity by amyloid-beta. Nature 2000; 403:98-103.
- 41. Oh, SP., Seki, T., Goss, KA., Imamura, T., Yi, Y., Donahoe, PK., Li, L., Miyazono, K., ten Dijke, P., Kim, S., and Li, E. Activin Receptor-Like Kinase 1 (ALK1) modulates TGF-b1 signaling in the regulation of angiogenesis. Proc Natl Acad Sci USA 2000 (in press).
- 42. Sado, T., Fenner, M. H., Tan, S.-S., Tam, P., Shioda, T., and Li, E. (2000). X inactivation in the mouse embryo deficient for Dnmt1: distinct effect of hypomethylation on imprinted and random X inactivation. *Dev. Biol.* 15: 294-303.
- 43. Kim, S.K., Hebrok, M., Li, E., Oh, S.P., Schrewe, H., Harmon, E.B., Lee, J.S., and Melton, D.A. (2000). Activin receptor patterning of foregut organogenesis. Genes Dev 14:1866-1871.
- 44. Pannell D, Osborne CS, Yao S, Sukonnik T, Pasceri P, Karaiskakis A, Okano M, Li E, Lipshitz HD, and Ellis J (2000) Retrovirus vector silencing is de novo methylase independent and marked by a repressive histone code. EMBO J. 19:5884-5894.
- 45. Sado, T., Wang, Z., Sasaki, H., and Li, E. (2001). Regulation of Imprinted X-Inactivation in Mice by Tsix. Development 128: 1275-1286.
- 46. Ferguson, C.A., Tucker, A.S., Heikinheimo, K., Nomura, M., Oh, P., Li, E., Sharpe, PT. (2001) The role of effectors of the activin signalling pathway, activin receptors IIA and IIB, and Smad2, in patterning of tooth. Development 128:4605-4613.
- 47. Ko, J., Humbert, S., Bronson, R.T., Takahashi, S., Kulkarni, A.B., Li, E., Tsai, L.H. (2001) p35 and p39 are essential for cyclin-dependent kinase 5 function during neurodevelopment. J Neurosci 21:6758-6771.
- 48. Jiang P, Song J, Gu G, Slonimsky E, Li E, Rosenthal N. (2002). Targeted deletion of the MLC1f/3f downstream enhancer results in precocious MLC expression and mesoderm ablation. Dev Biol 243:281-293.
- 49. Liang G, Chan MF, Tomigahara Y, Tsai YC, Gonzales FA, Li E, Laird PW, Jones PA. (2002). Cooperativity between DNA Methyltransferases in the Maintenance Methylation of Repetitive Elements. Mol Cell Biol 22(2):480-491.
- 50. Hata, H., Okano, M., Lei, H., and Li, E. (2002) Dnmt3L cooperates with the Dnmt3 family de novo DNA methyltransferases to establish maternal imprinting in mice. Development 129: 1983-1993.
- 51. Sakuma R, Ohnishi Yi Y, Meno C, Fujii H, Juan H, Takeuchi J, Ogura T, Li E, Miyazono K, Hamada H. (2002) Inhibition of Nodal signalling by Lefty mediated through interaction with common receptors and efficient diffusion. Genes Cells 7:401-412.
- 52. Miura K, Kishino T, Li E, Webber H, Dikkes P, Holmes GL, Wagstaff J. (2002) Neurobehavioral and electroencephalographic abnormalities in ube3a maternal-deficient mice. Neurobiol Dis 9:149-159.

- 53. Dodge, J. Ramsahoye, B.H., Wo, Z.G, Okano, M., and Li, E. (2002) *De novo* methylation of MMLV provirus in embryonic stem cells: CpG versus non-CpG methylation. Gene 289:41-48.
- 54. Oh. S.P., and Li, E. (2002). Gene-dosage sensitive genetic interactions between inversus viserum (iv), nodal, and activin type IIB receptor (ActRIIB) genes in asymmetrical patterning of the visceral organs along the left-right axis. Dev. Dyn. 224:279-290.
- 55. Chen T, Ueda Y, Xie S, Li E. (2002). A novel Dnmt3a isoform produced from an alternative promoter localizes to euchromatin and its expression correlates with active de novo methylation. J Biol Chem. 277(41):38746-54.
- 56. Oh, SP, Yeo, CY, Lee, Y, Schrewe, H, Whitman, M, and Li, E (2002). Activin type IIA and type IIB receptors mediate Gdfl1 signaling in axial vertebral patterning. Genes Dev. 16: 2749-2754.
- 57. Sado T, Li E, Sasaki H. Effect of TSIX disruption on XIST expression in male ES cells. Cytogenet Genome Res. 2002;99:115-118.
- 58. Trasler J, Deng L, Melnyk S, Pogribny I, Hiou-Tim F, Sibani S, Oakes C, Li E, James SJ, Rozen R. Impact of Dnmt1 deficiency, with and without low folate diets, on tumor numbers and DNA methylation in Min mice. Carcinogenesis. 2003 Jan;24(1):39-45.
- 59. Welte T, Zhang SS, Wang T, Zhang Z, Hesslein DG, Yin Z, Kano A, Iwamoto Y, Li E, Craft JE, Bothwell AL, Fikrig E, Koni PA, Flavell RA, Fu XY(2003). STAT3 deletion during hematopoiesis causes Crohn's disease-like pathogenesis and lethality: a critical role of STAT3 in innate immunity. Proc Natl Acad Sci U S A. 2003 Feb 18;100(4):1879-84.
- 60. Wei Chen, Yuqiong Liang, Wenjie Deng, Ken Shimizu, Amir M. Ashique, En Li, and Yi-Ping Li (2003). The zinc-finger protein CNBP is required for forebrain formation in the mouse. Development 130: 1367-1379.
- 61. Joost Gribnau, Konrad Hochedlinger, Ken Hata, En Li, and Rudolf Jaenisch (2003). Asynchronous replication timing of imprinted loci is independent of DNA methylation, but consistent with differential subnuclear localization. Genes Dev 17:759-773.
- 62. Lehnertz B, Ueda Y, Derijck AA, Braunschweig U, Perez-Burgos L, Kubicek S, Chen T, Li E, Jenuwein T, Peters AH. (2003) Suv39h-mediated histone h3 lysine 9 methylation directs DNA methylation to major satellite repeats at pericentric heterochromatin. Curr Biol. 13:1192-1200.
- 63. Genomic organization and promoter analysis of the Dnmt3b gene. (2003) Ishida C, Ura K, Hirao A, Sasaki H, Toyoda A, Sakaki Y, Niwa H, Li E, Kaneda Y. Gene. 310:151-9.
- 64. Taiping Chen, Yoshihide Ueda, Jonathan E. Dodge, Zhenjuan Wang, and En Li (2003) Establishment and Maintenance of Genomic Methylation Patterns in Mouse Embryonic Stem Cells by Dnmt3a and Dnmt3b. Mole. Cell Biol., 23:5594-5605.

- 65. Jacoby JJ, Kalinowski A, Liu MG, Zhang SS, Gao Q, Chai GX, Ji L, Iwamoto Y, Li E, Schneider M, Russell KS, Fu XY. Cardiomyocyte-restricted knockout of STAT3 results in higher sensitivity to inflammation, cardiac fibrosis, and heart failure with advanced age. Proc Natl Acad Sci U S A. 2003. 100:12929-34.
- 66. Kelly TL, Li E, Trasler JM. 5-aza-2'-deoxycytidine induces alterations in murine spermatogenesis and pregnancy outcome. J Androl. 2003. 24:822-830.
- 67. Mund C, Musch T, Strodicke M, Assmann B, Li E, Lyko F. Comparative analysis of DNA methylation patterns in transgenic Drosophila overexpressing mouse DNA methyltransferases. Biochem J. 2004. 378:763-768.
- 68. Sado T, Okano M, Li E, Sasaki H. De novo DNA methylation is dispensable for the initiation and propagation of X chromosome inactivation. Development. 2004. 131:975-982.
- 69. Dodge JE, Kang YK, Beppu H, Lei H, Li E. Histone H3-K9 methyltransferase ESET is essential for early development. Mol Cell Biol. 2004. 24:2478-86.
- 70. Kaneda M, Okano M, Hata K, Sado T, Tsujimoto N, Li E, Sasaki H. Essential role for de novo DNA methyltransferase Dnmt3a in paternal and maternal imprinting. Nature. 2004. 429(6994):900-3.
- 71. Beppu H, Ichinose F, Kawai N, Jones RC, Yu PB, Zapol WM, Miyazono K, Li E, Bloch KD. BMPR-II heterozygous mice have mild pulmonary hypertension and an impaired pulmonary vascular remodeling response to prolonged hypoxia. Am J Physiol Lung Cell Mol Physiol. 2004 Jul 30
- 72. de Sousa Lopes SM, Roelen BA, Monteiro RM, Emmens R, Lin HY, Li E, Lawson KA, Mummery CL. BMP signaling mediated by ALK2 in the visceral endoderm is necessary for the generation of primordial germ cells in the mouse embryo. Genes Dev. 2004 Aug 1;18(15):1838-49.

- 73. Hattori N, Abe T, Hattori N, Suzuki M, Matsuyama T, Yoshida S, Li E, Shiota K. Preference of DNA methyltransferases for CpG islands in mouse embryonic stem cells. Genome Res. 2004. 14(9):1733-40.
- 74. Chen T, Tsujimoto N, Li E. The PWWP Domain of Dnmt3a and Dnmt3b Is Required for Directing DNA Methylation to the Major Satellite Repeats at Pericentric Heterochromatin. Mol Cell Biol. 2004. 24:9048-58.
- 75. Fang J, Chen T, Chadwick B, Li E, Zhang Y. Ring1b-mediated H2A ubiquitination associates with inactive X chromosomes and is involved in initiation of X inactivation. J Biol Chem. 2004 Dec 17;279(51):52812-5.
- 76. Beppu H, Ichinose F, Kawai N, Jones RC, Yu PB, Zapol WM, Miyazono K, Li E, Bloch KD. BMPR-II heterozygous mice have mild pulmonary hypertension and an impaired pulmonary vascular remodeling response to prolonged hypoxia. Am J Physiol Lung Cell Mol Physiol. 2004 Dec;287(6):L1241-7.
- 77. Park S, Lee YJ, Lee HJ, Seki T, Hong KH, Park J, Beppu H, Lim IK, Yoon JW, Li E, Kim SJ, Oh SP. B-cell translocation gene 2 (Btg2) regulates vertebral patterning by modulating bone morphogenetic protein/smad signaling. Mol Cell Biol. 2004 Dec;24(23):10256-62.
- 78. Feng J, Chang H, Li E, Fan G. Dynamic expression of de novo DNA methyltransferases Dnmt3a and Dnmt3b in the central nervous system. J Neurosci Res. 2005 Mar 15;79(6):734-46.
- 79. Beppu H, Lei H, Bloch KD, Li E. Generation of a floxed allele of the mouse BMP type II receptor gene. Genesis. 2005 Feb 25;41(3):133-137
- 80. Yu PB, Beppu H, Kawai N, Li E, Bloch KD. Bone morphogenetic protein (BMP) type II receptor deletion reveals BMP ligand-specific gain of signaling in pulmonary artery smooth muscle cells. J Biol Chem. 2005. 280:24443-50.
- 81. <u>Hopfer U, Fukai N, Hopfer H, Wolf G, Joyce N, Li E, Olsen BR.</u> Targeted disruption of Col8a1 and Col8a2 genes in mice leads to anterior segment abnormalities in the eye. FASEB J. 2005. 19:1232-44.
- 82. Rodic N, Oka M, Hamazaki T, Murawski MR, Jorgensen M, Maatouk DM, Resnick JL, Li E, Terada N. DNA methylation is required for silencing of Ant4, an adenine nucleotide translocase selectively expressed in mouse embryonic stem cells and germ cells.

 Stem Cells. 2005 Jul 28; [Epub ahead of print]

Proceedings of Meetings

- 1. Li E, Beard C, Foster A, Bestor TH, Jaenisch R. DNA methylation, genomic imprinting, and mammalian development. Proceedings of the Cold Spring Harbor Symposium on Quantitative Biology; 1993; 58: 297-305.
- 2. Muragaki Y, Timmons S, Griffith CM, Oh SP, Li E, Fukai N, Fadel B, Quertermous T,

Olsen B. Tissue-specific expression of three alternative variants of *Col18*a1 and their localization in basement membrane zones. Proceedings of the 7th International Symposium on Basement Membranes, Bethesda, 1995.

3. Okano M, Li E. (2002). Genetic analyses of DNA methyltransferase genes in mouse model system. J Nutr. 132:2462S-5S.

Reviews, Book Chapters, and Editorials

- 1. Jaenisch R, Beard C, Li, E. DNA methylation and mammalian development. In: Ohlsson R, Hall K, Ritzen M, editors. Genomic imprinting: Causes and consequences. Cambridge, UK: Cambridge University Press; 1995. p. 118.
- 2. Li, E. Role of DNA methylation in mammalian development. In: Reik W, Sorani A, editors. Genomic Imprinting Frontiers in Molecular Biology. Volume 18. Oxford, UK: Oxford University Press; 1997. p. 1-20.
- 3. Li, E. The mojo of methylation [News & Views]. Nat Genet 1999; 23: 5-6.
- 4. Li, E and Jaenisch, R. DNA methylation and methyltransferases. In: Ehrlich M, editor. DNA alterations in cancer: Genetic and epigenetic changes. PP. 351-365. Natick: BioTechniques Books; 2000.
- 5. Li E. Chromatin modification and epigenetic reprogramming in mammalian development. 2002. Nature Rev. Genet. 3, 662-673.
- 6. <u>Chen T, Li E.</u> Structure and function of eukaryotic DNA methyltransferases. 2004. Curr Top Dev Biol. 60:55-89.

EXHIBIT B

EXHIBIT B

Alignment spanning the coding region of mouse Dnmt3a sequence from ATCC Deposit No. 209933 (top) and currently amended SEQ ID NO:1 (bottom)¹

atgccctccagcggccccggggacaccagcagctcctctctggagcgggaggatgatcga 	276
aaggaaggaggaacaggaggagaaccgtggcaaggaagagcgccaggagcccagcgcc 	336
acggcccggaaggtggggaggcctggccggaagcgcaagcacccacc	396
gacacccccaaggacccagcagtgaccaccaagtctcagcccatggcccaggactctggc	456
ccctcagatctgctacccaatggagacttggagaagcggagtgaaccccaacctgaggag	516
gggagcccagctgcagggcagaagggtggggccccagctgaaggagagggaactgagacc 	576
ccaccagaagcctccagagctgtggagaatggctgctgtgtgaccaaggaagg	636
gcctctgcaggagagggcaaagaacagaagcagaccaacatcgaatccatgaaaatggag 	696
ggctcccggggccgactgcgaggtggcttgggctgggagtccagcctccgtcagcgaccc 	756
atgccaagactcaccttccaggcagggacccctactacatcagcaaacggaaacgggat 	816
gagtggctggcacgttggaaaagggaggctgagaagaaagccaaggtaattgcagtaatg 	876
aatgctgtggaagagaaccaggcctctggagagtctcagaaggtggaggaggccagccct 	936
cctgctgtgcagcagcccacggaccctgcttctccgactgtggccaccacccctgagcca	996

¹ Bolded nucleotides indicate nucleotides that were amended on July 23, 2001.

gtaggagggatgctggggacaagaatgctaccaaagca g ccgacgatgagcctgagtat 	, 056
gaggatggccggggctttggcattggagagctggtgtgggggaaacttcgggggcttctcc 	116
tggt [:] ggccaggccgaattgtgtcttggtggatgacaggccggagccgagcagctgaaggc 	176
actcgctgggtcatgtggttcggagatggcaagttctcagtggtgtgtgt	236
atgeegetgageteettetgeagtgeatteeaceaggeeacetacaacaageageeeatg 	296
taccgcaaagccatctacgaagtcctccaggtggccagcagccgtgccgggaagctgttt 	356
ccagcttgccatgacagtgatgaaagtgacagtggcaaggctgtggaagtgcagaacaag 	116
cagatgattgaatgggccctcggtggcttccagccctcgggtcctaagggcctggagcca 	176
ccagaagaagaagaatccttacaaggaagtttacaccgacatgtgggtgg	36
gcagctgcttacgcccaccccaccagccaagaaacccagaaagagcacaaca	96
cctaaggtcaaggagatcattgatgagcgcacaagggagcggctggtgtatgaggtgcgc 	56
cagaagtgcagaaacatcgaggacatttgtatctcatgtgggagcctcaatgtcaccctg 	16
gagcacccactcttcattggaggcatgtgccagaactgtaagaactgcttcttggagtgt 	76
gcttaccagtatgacgacgatgggtaccagtcctattgcaccatctgctgtgggggggg	36
aagtgeteatgtgtgggaacaacaactgetgeaggtgettttgtgtegagtgtgtggat 	96

ctcttggtggggccaggagctgctcaggcagccattaaggaagacccctggaactgctac	6
atgtgcgggcataagggcacctatgggctgctgcgaagacgggaagactggccttctcga	б
ctccagatgttctttgccaataaccatgaccaggaatttgaccccccaaaggtttaccca	
cctgtgccagctgagaagaggaagcccatccgcgtgctgtctctctttgatgggattgct	5
acagggctcctggtgctgaaggacctgggcatccaagtggaccgctacattgcctccgag	5 ·
gtgtgtgaggactccatcacggtgggcatggtgcggcaccagggaaagatcatgtacgtc	5
ggggacgtccgcagcgtcacacagaagcatatccaggagtggggcccattcgacctggtg	5
attggaggcagtccctgcaatgacctctccattgtcaaccctgcccgcaagggactttat	5
gagggtactggccgcctcttctttgagttctaccgcctcctgcatgatgcgcggcccaag	5
gagggagatgatcgccccttcttctggctctttgagaatgtggtggccatgggcgttagt	5
gacaagagggacatctcgcgatttcttgagtctaaccccgtgatgattgacgccaaagaa	5
gtgtctgctgcacacagggcccgttacttctggggtaaccttcctggcatgaacaggcct	į
ttggcatccactgtgaatgataagctggagctgcaagagtgtctggagcacggcagaata	;
gccaagttcagcaaagtgaggaccattaccaccaggtcaaactctataaagcagggcaaa	
gaccagcatttccccgtcttcatgaacgagaaggaggacatcctgtggtgcactgaaatg	

gaaagggtgtttggcttccccgtccactacacagacgtctccaacatgagccgcttggcg .	2856
aggcagagactgctgggccgatcgtggagcgtgccggtcatccgccacctcttcgctccg	2916
ctgaaggaatattttgcttgtgtgtaagggacatgggggcaaactgaagtagtgatgata 	2976
aaaaagttaaacaaacaaacaccaagaacgagaggacggagaaaagttcagcaccc	3037
agaagagaaaaaggaatttaaagcaaaccacagaggaggaaaacgccggagggcttggcc	3098
ttgcaaaagggttggacatcatctcctgagttttcaatgttaaccttcagtcctatctaa	3158
aaagcaaaataggccctccccttcttcccctccggtcctaggaggcgaactttttgttt	3218
tctactctttttcagaggggttttctgtttgtttgggtttttgtttcttgctgtgactga	3278
aacaagagagttattgcagcaaaatcagtaacaacaaaaagtagaaatgccttggagagg	3338
aaagggagagagggaaaattctataaaaacttaaaatattggtttttttt	3398
ttctatatatctctttggttgtctctagcctgatcagataggagcacaaacaggaagaga	3458
atagagaccctcggaggcagagtctcctctcccacccccgagcagtctcaacagcacca	3518
ttcctggtcatgcaaaacagaacccaactagcagcagggcgctgagagaacaccacaca	3578
gacactttctacagtatttcaggtgcctaccacacaggaaaccttgaagaaaaccagttt	3638
ctagaagccgctgttacctcttgtttacagtt 	

EXHIBIT C

EXHIBIT C

Alignment spanning the coding region of mouse Dnmt3b from ATCC Deposit No. 209934 (top) and currently amended SEQ ID NO:2 (bottom)²

caggaaacaatgaagggagacagcagacatctgaatgaagaagaggtgccagcgggtat caggaaacaatgaagggagacagcagacatctgaatgaagaagaggtgccagcgggtat 3	319
gaggagtgcattatcgttaatgggaacttcagtgaccagtcctcagacacgaaggatgct	379
ccctcaccccagtcttggaggcaatctgcacagagccagtctgcacaccagagaccaga	139
ggccgcaggtcaagctcccggctgtctaagagggaggtctccagccttctgaattacacg	199
caggacatgacaggagatggagacagagatgatgaagtagatgatgggaatggctctgat 	559
attctaatgccaaagctcacccgtgagaccaaggacaccaggacgcgctctgaaagcccg 	19
gctgtccgaacccgacatagcaatgggacctccagcttggagaggcaaagagcctccccc 	79
agaatcacccgaggtcggcagggccgccaccatgtgcaggagtaccctgtggagtttccg 	39
gctaccaggtctcggagacgtcgagcatcgtcttcagcaagca	99
gccagcgtcgacttcatggaagaagtgacacctaagagcgtcagtaccccatcagttgac 	59
ttgagccaggatggagatcaggagggtatggataccacacaggtggatgcagagagca g a 	19
gatggagacagcacagagtatcaggatgataaagagtttggaataggtgacctcgtgtgg 	79
ggaaagatcaagggcttctcctggtggcctgccatggtggtgtcctggaaagccacctcc	
	039

² Bolded nucleotides indicate nucleotides that were amended on July 23, 2001.

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aagcgacaggccatgcccggaatgcgctgggtacagtggtttggtgatggcaagttttct	1099
gagatetetgetgacaaactggtggetetggggetgtteageeageactttaatetgget	1159
accttcaataagctggtttcttataggaaggccatgtaccacactctggagaaagccagg	1219
gttcgagctggcaagaccttctccagcagtcctggagagtcactggaggaccagctgaag	1279
cccatgctggagtgggcccacggtggcttcaagcctactgggatcgagggcctcaaaccc	1339
aacaagaagcaaccagtggttaataagtcgaaggtgcgtcgttcagacagtaggaactta	1399
gaacccaggagacgcgagaacaaaagtcgaagacgcacaaccaatgactctgctgcttct	1459
gagtccccccacccaagcgcctcaagacaaatagctatggcgggaaggaccgaggggag	1519
gatgaggagagccgagaacggatggcttctgaagtcaccaacaacaagggcaatctggaa	L579
gaccgctgtttgtcctgtggaaagaagaaccctgtgtccttccaccccctctttgagggt	L639
gggctctgtcagagttgccgggatcgcttcctagagctcttctacatgtatgatgaggac	L699
ggctatcagtcctactgcaccgtgtgctgtgagggccgtgaactgctgctgtgcagtaac	.759
acaagctgctgcagatgcttctgtgtggagtgtctggaggtgctggtgggcgcaggcaca	.819
gctgaggatgccaagctgcaggaaccctggagctgctatatgtgcctccctc	.879
catggggtcctccgacgcaggaaagattggaacatgcgcctgcaagacttcttcactact	939

gatcctgacctggaagaatttgagccacccaagttgtacccagcaattcctgcagccaaa	
aggaggcccattagagtcctgtctctgtttgatggaattgcaacggggtacttggtgctc	2059
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agattcctggcatgtaacccagtgatgatcgatgccatcaaggtgtctgctgctcacagg	2479
gcccggtacttctggggtaacctacccggaatgaacaggcccgtgatggcttcaaagaat 	2539
gataagctcgagctgcaggactgcctggagttcagtaggacagcaaagttaaagaaag	2599
cagacaataaccaccaagtcgaactccatcagacagggcaaaaaccagcttttccctgta 	2659
gtcatgaatggcaaggacgacgttttgtggtgcactgagctcgaaaggatcttcggcttc 	2719
cctgctcactacacggacgtgtccaacatgggccgcggcgccgtcagaagctgctgggc 	2779
aggtcctggagtgtaccggtcatcagacacctgtttgcccccttgaaggactactttgcc 	2839

Li et al. Appl. No. 09/720,086

EXHIBIT D

LETTER

Estimation of Errors in "Raw" DNA Sequences: A Validation Study

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As DNA sequencing is performed more and more in a mass-production-like manner, efficient quality control measures become increasingly important for process control, but so also does the ability to compare different methods and projects. One of the fundamental quality measures in sequencing projects is the position-specific error probability at all bases in each individual sequence. Accurate prediction of base-specific error rates from "raw" sequence data would allow immediate quality control as well as benchmarking different methods and projects while avoiding the inefficiencies and time delays associated with resequencing and assessments after "finishing" a sequence. The program PHRED provides base-specific quality scores that are logarythmically related to error probabilities. This study assessed the accuracy of PHRED's error-rate prediction by analyzing sequencing projects from six different large-scale sequencing laboratories. All projects used four-color fluorescent sequencing, but the sequencing methods used varied widely between the different projects. The results indicate that the error-rate predictions such as those given by PHRED can be highly accurate for a large variety of different sequencing methods as well as over a wide range of sequence quality.

In DNA sequencing, knowledge about the accuracy of sequences can be very valuable. For example, different large scale sequencing projects may produce sequences at similar rates and costs but with significantly different error rates in the final sequence. One major determinant in the final error rate is the accuracy of the "raw" sequence. Knowledge about the frequency and location of errors in the raw sequence data can help to direct "polishing" efforts to the places where additional effort is needed; it also enables the comparison between different sequencing projects without requiring that the same region be sequenced in each project.

Another area where estimates about sequence error rates would be beneficial is technology development. Accurate error estimates at each base would enable "quality benchmarking" between different methods, thus enabling researchers to choose the method that fills their needs for accuracy and throughput best.

Several groups have developed mathematical models to predict the error probability at any given position in raw sequences. Lawrence and Solovyev used linear discriminant analysis to calculate separate probability estimates for insertions, deletions, and mismatches (Lawrence and Solovyev 1994). Ewing and Green (1998) developed the program

PHRED, which calculates a quality score at each base. This quality score q is logarithmically linked to the error probability p: $q = -10 \times 100_{10}$ (p) (for a discussion of how quality scores are calculated and what the limitations are, see Ewing et al. (1998). When used in combination with sequence assembly and finishing programs that utilize these error estimates, reliable error probabilities promise to increase the accuracy of consensus sequences and to reduce the efforts required in the finishing phase of sequencing projects (Churchill and Waterman 1992; Bonfield and Staden 1995).

To examine the accuracy of probability estimates made by the program PHRED, we compared the actual and predicted error rates for six different cosmid- or BAC-sized projects that were produced by six different large-scale sequencing centers in the United States. All of these six projects used four-color fluorescent sequencing machines; however, the DNA preparation methods, sequencing enzymes, fluorescent dyes and chemistries, and gellengths varied significantly between the six groups. Table 1 gives an overview of the sequencing projects analyzed. Table 2 lists the different methods used.

RESULTS

Error Rate Prediction Accuracy for Six Projects

A comparison of actual and predicted error rates for the six projects in this study is shown in Table 3.

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			Average	
Project	Reads	Aligned bases	allgned read length	
A	455	416,214	915	
В	1277	871,230	682	
C	1065	603,655	567	
D	834	414,595	497	
Ē	1638	1,149,209	702	
F	1885	907,796	482	
Total	7154	4,362,699	610	

The results indicate that PHRED is very successful in identifying bases with low error probabilities. For example, the 1.28 million bases with quality scores of 4–12 (corresponding to error probabilities between 39.8% and 6.3%) contain a total of 187,926 errors. In contrast, the 1.44 million bases with quality scores between 33 and 42 (corresponding to error probabilities between 0.05% and 0.006%) contain only 237 errors, which translates into a 790-fold lower error rate. The trend toward lower error rates can also be observed for each individual project. In most cases, the actual number of errors is close to the predicted error rate. It is also apparent that the actual error rate is typically lower than the predicted error rate.

Both the high overall accuracy and the tendency to slightly overpredict errors are confirmed by statistical analysis, as shown in Table 4. The correlation between predicted and actual error frequencies is excellent for all projects (Spearman correlation coefficient >0.89, P < 0.0001). Averaged over all projects, the actual error rate is 84.5% of the predicted error rate; the slope of the relation between predicted and actual error rates differs slightly between projects and ranges from 76.6% to 88.4%. To put these differences between projects in relation, it is worthwhile remembering that PITRED quality scores cover a wide dynamic range: The maximum quality score of 51 corresponds to a 50,000-fold lower predicted error rate than the minimum quality score of 4. Even the relative difference between successive quality is larger than the relative difference in the slopes; for example, a quality score of 10 corresponds to an error probability of 10%, whereas a score of 9 corresponds to an error probability of 12.6%.

A different way of looking at the relation between the actual and predicted error rates is shown in Figure 1. Here, the error rates as a function of the position within all reads in each of the projects, averaged over 50-base windows, is depicted. For all six projects, the predicted error rates are very close to the actual error rates over the entire length of the sequences. Each project has a characteristic distribution of error rates, which differs from each of the other projects. The minimum error rate differs dramatically between projects. The best projects achieve raw error rates of 0.23%—0.36% in the best region of the sequence read, typically from base 150 to 200. The worst project in the data set had an ~10-fold higher error rate of 2.58%.

Toward the end of sequence reads, the error rates increase and start to exceed 10% between bases 300 and 700. In projects that used mainly short gels (e.g., projects D and F), this increase begins sooner, whereas projects that use longer gels show a markedly longer stretch of low error rates (e.g., projects Λ and B).

Table 5 summarizes key results for the six projects. The first four projects have similar minimum and average error rates. However, the length of the region where the error rate is below 5% differs significantly, from 403 to 682 bases. The project with the shorter low error rate regions contained larger portions of reads generated on short gels, whereas projects A and B were run exclusively on long gels (ABI373 stretch or ABI377 sequencers). Other factors contributing to differences between the first four projects were differences in sequencing chemistries, production scale, and electrophoresis conditions and machines.

Project E and, in particular, project F, had significantly higher error rates than the first four projects. In projects E and F, every sequence generated for the project had been included in the data sct, whereas the other four projects had eliminated some "bad" sequences through manual or auto-

Table 2. Overview of Sequencing Methods Used in the Different Projects

Template DNA	single-stranded M13,
	double-stranded plasmids
Sequencing	Sequenase, Tag, KlenTagTR,
enzymes	AmpliTag FS
Sequencing	Dyes primer (two different dyes
chemistries	chemistries), dye terminalor
Sequencing	ABI 373, ABI 373 stretch,
machines	ABI 377
Gel length	Only short gels, only long gels.
	mixes of short and long gels

Table 3. Comparison of Predicted and Actual Error Rates for 5ix Different Sequencing Projects								
Project	Quality score	4-12	13-22	23–32	33-42	43-51		
A	aligned bases	119,246	75,293	70,391	144,876	73,234		
	expected errors	20,256	2,064	172	37	1		
	actual errors	16,784	1,758	127	17	1		
В	aligned bases	182,034	137,940	181,998	399,690	140,176		
	expected errors	29,953	3,704	410	102	3		
	actual errors	26,038	2,536	287	35	0		
Ċ	aligned bases	139,345	131,419	151,197	292,070	68,529		
	expected errors	22,277	3,411	35 <i>7</i>	74	2		
	actual errors	16,670	1,513	194	26	3		
D	aligned bases	103,898	68,995	68,613	153,730	111,752		
	expected errors	16,880	1,919	168	38	3		
	actual errors	14,495	1,924	146	59	2		
E	aligned bases	378,755	217,438	167,968	392,717	144,313		
	expected errors	63,947	6,336	418	95	4		
	actual errors	55,968	6,516	355	67	5		
F .	aligned bases	359,809	136,688	98,840	64,035	5,1:30		
	expected errors	66,938	4,079	256	23	0		
	actual errors	57,971	3,856	332	33	1		
All	aligned bases	1,283,087	767,773	739,007	1,447,118	543,134		
	expected errors	220,252	21,513	1,781	370	13		
	actual errors	187,926	18,103	1,441	237	12		

matic inspection. After eliminating <10% of the worst sequences in project F, the error rates for the remaining sequences were comparable to those of the first four projects. In contrast, project F showed a much more uniform distribution of sequence quality.

The last column in Table 5 shows the average number of bases with an estimated error probability of at most 0.1%, which is equivalent to a quality score of at least 30. The count of such "very high-quality" bases is a good indicator of sequence quality, both for individual sequences and, when aver-

Table 4.	Summary of Stat	istical Analysis	Results		
Project	Spearman p	P > p	Slope	t ratio	P > t
A	0.9646	<0.0001	0.818	75.1	<0.0001
В	0.9890	< 0.0001	0.874	98.2	< 0.0001
C	0.9846	< 0.0001	0.766	71.6	< 0.0001
D•	0.8692	< 0.0001	0.855	68.3	< 0.0001
∴E'	0.9956	< 0.0001	0.884	144.3	<0.0001
F	0.9968	< 0.0001	0.865	151.6	< 0.0001
All	0.9964	< 0.0001	0.845	174.5	< 0.0001

In project D, the Spearman correlation coefficient p was artificially low as only very few bases (10) bases had a quality score of 5, and none of these bases contained an actual error (expected: 3.16 errors). Exclusion of this quality score gave a Spearman correlation coefficient of 0.9786 (P < 0.0001). The frequencies in the slope calculations were weighed by the number of bases at any given quality score and, thus, were not sensitive to such small sample distortions (see Methods).

ity analysis and control in large-scale DNA sequencing projects. To analyze how accurate PHRED error estimates are for different quality sequences within the same sequencing project, we subdivided a data set into four quartiles, based on the number of very high-quality bases in each sequence (see Methods). The comparison of actual and predicted error rates is shown in Figure 2.

When measured by the error rate in the best region of a sequence, the data quality in the different quartiles varies >100 fold between the best and the worst 25% of the sequences. The best quartile showed <0.03% error for >100 bases, whereas the error rate in the worst quartile always exceeded 5%. In quartiles 2 and 3, the predicted error rates match the actual error rates very closely. In the best and

worst quartiles, PHRED's accuracy was somewhat lower from base 100 to 500. In the best sequences, PHRED's error estimates were about twofold too high; in the worst sequences, the error estimates were too low, again by a factor of 2. This underprediction of errors can be partially explained by the fact that PHRED gives ambiguous base calls (N's) a quality score of 4, corresponding to an error probability of 39.8%; however, N's will always show up as an actual error. Even in the worst and best quartiles, however, the predicted error rate curves are very similar to the actual error rate curves.

The results shown in Figure 2 also demonstrate that the count of very high-quality bases, or bases with an estimated error probability of at most 0.1%, can be used effectively to characterize the overall

quality of a sequence read. Sorting the sequence reads into quartiles based on the number of very high-quality bases worked well, as shown by the >100-fold difference in the minimum error rate between the first and the fourth quartile.

Other methods to characterize the overall quality of individual reads based on PHRED quality scores can give similar results. For example, counting bases above a minimum quality threshold anywhere in the range of 20-40 gave similar results for most data sets (not shown), and such counts are used by a number of different laboratories as quality measures. Alternatively, the quality values can be converted to error probabilities and averaged to give the predicted error rate for the trace, or summed to give the total predicted number of errors in a trace. However, such averages and totals can sometimes give a misleading picture, as the following example illustrates. Assume that two sequence reads have very similar quality in the alignable part of the read but that one of the two sequences was run much longer and

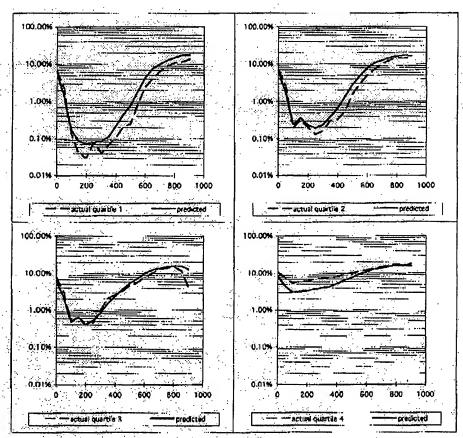


Figure 2 Actual and predicted error rates in different quality subsets of project B. Sequence reads were sorted by the number of bases with a predicted error rate of at most 0.1% (very high-quality bases), and assigned to quartiles, with quartile 1 corresponding to the highest numbers. Actual and predicted error rates for all sequences in each subset were calculated as in Fig. 1. Note that a number of sequence reads that had been rejected because of too low quality were added back to the data set for illustrative purposes, all of which are in quartile 4. These sequences were not included in the data sets used to generate Figs. 1 and 3 and Tables 1 and 3.

RICHTERICH

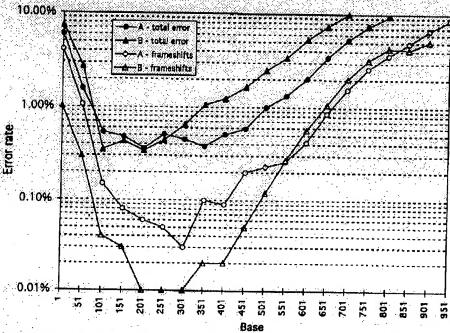


Figure 3 Actual frameshift and total error rates for projects A and B. To calculate frameshift error rates, only insertions and deletions were counted. Mismatch errors, which account for the vast majority of errors after base 150, were included only in the total error count. Note that project B (\triangle , \triangle) has a slightly similar or slightly higher total error rate compared to project A (\bullet ,C) but only about one-third as many insertions and deletions up to base 500. For both projects, the frameshift error rate in the raw data is <1 in 1000 for >300 bases, and \le 1 in 10,000 for >100 bases in project B.

therefore contains a longer unalignable "tail" of very low-quality bases. When calculating the average error rate for these two sequences, the second sequence will have a much higher average error and, therefore, appear to be of lower quality. In contrast, the counts of very high-quality bases for both sequences will be very similar, as the unalignable tails contain few, if any, high-quality bases. Therefore, counts of bases above a high chough quality threshold will give a more robust and clearer picture of trace quality.

Frameshift Error Rates for Different Sequencing Chemistries

Depending on how biologists use DNA sequences, knowledge about total error rates in raw sequences may or may not be sufficient. For example, frame-shift errors in coding sequences will generally lead to incorrectly predicted open reading frame, whereas mismatch errors will do so only if the mismatch introduces a stop codon or a new splice site. At the time of this writing, PHRED did not differentiate between mismatch and frameshift errors, but only estimated total error rates. This might occa-

sionally lead to questionable conclusions, as the results shown in Figure 3illustrate.

Figure 3 shows the total actual error rates and the frameshift error rates for two projects, A and B. The total error rates for both projects are similar for up to 350 hases; after 350 bases, project B has a somewhat higher total error rate. However, examining the frameshift error rate gives rise to a different picture: from base 1 to 500, project A has approximately four times as many insertions and deletions as project B. This difference in frameshift error rates can be explained by the sequencing chemistries that were used in the two projects. Project B, with the lower frameshift error rate, used only dye terminator chemistry, which is known to eliminate band spacing artifacts from hairpin structures ("compressions"). Project A, on the

other hand, used dye primer chemistry, which is more prone to insertion and deletion errors from mobility artifacts, for most sequencing reactions.

DISCUSSION

As large-scale DNA sequencing has become a more routine and common process, the traditional methods for assessing sequence quality have become unsatisfactory. In projects like single-pass cl)NA sequencing, it is not possible to calculate and compare error rates after finishing a sequence, as finishing never takes place. Even when a comparison between raw and finished sequence can be done, the time delay between raw data generation and quality assessment is often large. This delay makes it difficult to improve ongoing projects, and it sometimes makes it impossible to capture problems early on. Some immediate quality feedback can be reached by including known standard sequences for quality control. However, this approach can be costly, and it fails when error profiles differ between standard and unknown sequences.

In contrast to these traditional methods to assess sequence accuracy, direct estimation of error rates in raw sequence data would enable immediate quality control and feedback. Accurate, base-by-base estimates of error probabilities could also increase the utility of single-pass sequences significantly, allow efficient comparison and optimization of different sequence chemistries, and enable the development of better software tools for sequence assembly and analysis.

The critical question for any error rate prediction tool is how accurate are the error rate estimates, in particular if different sequencing methods and chemistries are used? The results presented herein provide an answer to this question for the program PHRED, as well as clues where further development would be useful. As shown in Tables 3 and 4 and in Figure 1, the agreement between predicted and actual error rates was very good in each of the six different projects analyzed. The observed high level of prediction accuracy in all of these projects is almost astonishing if one takes into account that actual errors are binary (a base is either correct or wrong), whereas predicted error rates are probabilities on a scale from 0.0 to 1.0. The observed tendency to overpredict error rates can be at least partially explained by the "small sample correction" that was used in the derivation of threshold parameters for quality scores (liwing and Green 1998). For most practical applications, such a somewhat conservative estimation of quality scores is tolerable or even desirable. Overall, the results clearly show that error probabilities given by PHRED accurately describe raw sequence data quality.

In Judging the usefulness of predicted error probabilities, it is important to know how differences in sequencing methods will influence the prediction accuracy. For example, the larger variation in peak heights tends to be larger in dye terminator sequencing than in dye primer sequencing, and different sequencing enzymes are known to produce different specific height variation patterns. Any estimation of error probabilities that takes the peculiarities of a specific sequencing chemistry into account would therefore be expected to be less accurate for different chemistries.

The projects included in this study were specifically chosen to provide an initial answer to the question of how generally useful PHRED quality scores are. These projects represent the vast majority of different multicolor fluorescent sequencing methods used in the last 3 years: different template DNAs and DNA preparation methods, different enzymes, gel lengths, run conditions, and different fluorescent dyes. The data also include a considerable spread in data quality, both between projects

and within individual projects. None of the projects analyzed here were included in PHRED's training sel, and just one of the six laboratories that contributed data to this study also contributed data to the training data sets. One of the projects in this sludy consisted entirely of dye terminator sequences, which presented only a small fraction of the sequences in the test data set. Another project exclusively used a set of fluorescent dyes different from those used in the training sets. Fach project differed from the other projects in this study in at least one, and typically many, experimental aspects like template preparation, sequencing enzymes, gel run conditions, and so forth. Despite these differences, the accuracy of error rate predictions was very similar for all projects.

Our results justify some optimism about the accuracy of PHRED quality scores for minor changes in sequencing technology, for example, sequences generated by new enzymes and fluorescent dyes. Initial studies showed that PHRED quality scores were also accurate for sequences produced by multiplex sequencing with radioactive detection (P. Richterich, unpubl.). However, we also observed two effects that can invalidate PHRED quality scores during these studies. First, sequences generated by chemical sequencing gave too low quality scores at mixed (A + G) reactions. Because secondary peak height is one of the parameters used in the error rate predictions, this is not surprising. Another potential source of error is high-frequency noise in the trace data. With such data, PHRED occasionally underestimated the hand spacing by a factor of 2 or more, which resulted in incorrect base calls and quality scores. By applying simple smoothing algorithms to data with high-frequency noise, these problems could typically be resolved. Similar steps may be necessary to obtain accurate PHRED quality scores on data that have been generated by different sequencing instruments or preprocessed by different software.

Accurate quality scores can have a major impact on how sequences are used downstream from the sequence production process. In traditional sequencing projects where the goal is complete coverage at a final error rate below (e.g.) I in 10,000, the accuracy goals can be reached with single sequence reads as long as the quality scores are at least 40 (however, other potential problems like clone instability may make higher coverage advisable). Interesting questions arise as to how individual read quality contributes to project quality, or the error rate of the "final" sequence. Under the assumption that errors between different sequence reads are

completely independent, one could argue that two reads with a quality score of 20 (error probability of I in 100) are just as valuable as one sequence with a quality score of 40 (error probability of 1 in 10,000). However, although a single sequence stretch with quality levels above 40 would give a final sequence with an error rate of <1 in 10,000, assembling a consensus from two sequences with quality scores of 20 (1% error rate) could lead to one of two results: If the errors were completely random, the consensus sequence would be ambiguous at 2% of all locations; if the errors were completely localized, for example, because of reproducible compressions, the consensus sequence would have one "hidden" error every 100 bases. Typically, consensus sequences derived from low-quality sequences will have both kinds of problematic regions. Increased coverage can rapidly eliminate the random errors; however, increased coverage does not resolve errors from systematic sources. Manual examination of such problem areas is generally required; such "contig editing," however, tends to be time consuming, requires highly trained personnel, is an obstacle toward complete automation of DNA sequencing, and sometimes fails to climinate all errors. This leads to the somewhat counterintuitive conclusion that the practical value of increasing sequence qual-Ity can be even higher than indicated by the quality scores: One sequence of average quality above 40 can be "worth" more than two sequences of average quality 20.

Another application of DNA sequencing where high quality can be of disproportionately high value is the search for mutations in genomic DNA. In low quality sequences, secondary peaks and low resolution often complicate the identification of heterozygous mutations. In regions of higher sequence quality, such secondary peaks are smaller or absent and peaks are better resolved. Therefore, both falsepositive and false-negative errors can be significantly reduced in high-quality regions. Tools like PHRED, which can accurately measure sequence quality from trace data, can he of twofold value for mutation detection. First, base-specific quality scores can allow optimization of sequencing methods and strategies for mutation detection. Second, the quality scores can be used to evaluate the usefulness of individual sequence reads for mutation detection (e.g., by discarding reads below minimum thresholds), and they can guide software that automatically detects mutations.

The ability to predict error rates in a highly accurate fashion is likely to have a major impact in applications like those described above. PHRED is

the first widely used program that accurately predicts base-specific error probabilities. However, the algorithm for determining quality values has been described (Ewing and Green 1998), and it should be straightforward to implement similar quality values in other base-calling programs. Furthermore, an extension of the approach developed by kwing and Green should be possible. For example, differentiation between mismatch and frameshift errors would enable better comparisons of sequencing methods with similar total error rates but different frameshift error rates. Several groups have described efforts to calculate separate probabilities (or "confidence assessments") for mismatch errors and frameshift errors (Lawrence and Solovyev 1994; Berno 1996). Their results demonstrated that different approaches to error type characterization are feasible and promising. Implementation of such error type predictions in other programs similar to the way PHRED uses quality scores would enable better method assessments, benchmarking, and production quality control, and could have a significant impact on downstream uses of DNA sequence information.

METHODS

Data Sets

For one project, sequence raw data in the form of ABI trace files were downloaded from a public FTP site. Sequence data for the five other projects were kindly provided by five different large-scale sequencing groups. Table I gives a summary of the six projects, and Table 2 gives an overview of the different sequencing methods used in the projects. The projects differed in the amount of prescreening of data that had been done, reflecting different approaches to quality control in different laboratories. In two projects (B and C), different software programs had been used to identify and eliminate low-quality sequences. One project (F) included all data files generated, whereas the other three projects had excluded "falled lanes."

Comparison of Actual and Predicted Error Rates

The sequences for all traces in each project were recalled using the program PHRED (v. 961028). Next, sequences in each project were assembled with PHRAP (P. Green, unpubl.). Slightly different methods were chosen for the statistical and graphical evaluation of the error rate prediction accuracy. In the statistical evaluation, only the longest contig produced by PHRAP was considered. The tables of aligned bases and observed discrepancy counts for

each quality score were taken from the PHRAP output and analyzed as follows. The expected number of discrepancies (E) at each quality score (u) was calculated by multiplying the number of aligned bases (N) with the error probability corresponding to the quality score: $E = N \cdot 10^{-0.14}$. The Spearman ranking coefficients were calculated by comparing the expected and observed error frequencies. To obtain the quantitative relation between the expected and observed error rates over the entire range, a least-squares fit between the observed and expected rates was performed, with the intercept set to zero and the number of aligned bases at each quality score used as weights:

For a graphical comparison of estimated and actual error rates in 50-bp windows, the following steps were taken. For two of the projects, the consensus sequence was retrieved from public databases. For the four other projects, the DNA sequence and quality information were used by the program PHRAP to assemble consensus sequences for each of the projects. The individual reads were aligned to the consensus sequences of the longest config, using the program CROSS_MATCH (P. Green, unpubl.), after removing single-coverage regions from the ends of the consensus sequence. CROSS-

MATCH uses an implementation of the Smith Waterman algorithm to generate alignments that typically do not include the ends of sequences, where disagreements are commonly due to vector sequence or low quality sequence.

The quality files generated by PHRED and the alignment summaries generated by CROSS-MATCH were then analyzed as follows. First, the region of each query sequence that had been aligned by CROSS MATCH was determined. Next, the actual

by CROSS_MATCH was determined. Next, the actual and predicted error rates for the entire aligned part of each individual sequence was calculated. In addition, the average actual and predicted error rates for all alignable sequences together were calculated for windows of 50 bases in length. To calculate the predicted error rate, the quality scores q determined by PHRED at each base were converted to error probabilities as described above (Ewing and Green 1998).

Subdividing Data into Subsets Based on Data Quality

To examine the accuracy of PHRED quality scores for data subsets of different quality within a project, the following approach was taken. For all sequence reads in project B, the number of bases with a quality score of at least 30 in each sequence was determined (bases with quality scores of at least 30 were called very high-quality bases, or VHQ bases). Se-

quences were sorted in descending order based on the number of very high-quality bases, and divided into four quartiles. Accordingly, quartile 1 contained 25% of sequences with the highest number of very high-quality bases, and quartile 4 contained the "worst" sequences. To illustrate the prediction accuracy in data with relatively high error rates, sequences from project B that had been "discarded" because they had not met the minimum quality criteria were added back to the data set. The sequences in each quartile were compared to the consensus sequences that had been generated using the entire data set, as described above for the graphical comparison.

Determining Actual Frameshift Error Rates

The calculation of actual frameshift error rates in the raw sequence data was performed using CROSS MATCH, similar to the procedure described above for total error rates, except that only insertion and deletion errors were counted. Because PHRED does not give separate frameshift error estimates, a comparison of predicted and actual frameshift errors is not possible.

ACKNOWLEDGMENTS

I thank the participating laboratories for contributing their data; Dr. Josee Dupuis for help with the statistical analysis, and Dr. Phil Green for helpful discussions.

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REFERENCES

Berno, A.J. 1996. A graph theoretic approach to the analysis of DNA sequencing data. Genome Res. 6: 80-91.

Bonfield, J.K. and R. Staden. 1995. The application of numerical estimates of base calling accuracy to DNA sequencing projects. Nucleic Acids Res. 23: 1406-1410.

Churchill, G. and M.S. Waterman. 1992. The accuracy of DNA sequences: estimating sequence quality. *Genomics* 14: 89-98.

Ewing, B. and P. Green. 1998. Base-calling of automated sequencer traces using *phred*. II. Error probabilities. *Genome Res.* (this issue).

Ewing, B., L. Hillier, M.C. Wondl, and P. Green. 1998. Base-calling of automated sequencer traces using plired. 1. Accuracy assessment. Genome Res. (this issue).

Lawrence, C.B. and V.V. Solovyev. 1994. Assignment of position-specific error probability to primary sequence data. *Nucleic Acids Res.* 22: 1272–1280.

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IN THE UNITED STATES PATENT AND TRADEMARK OFFICE



In re application of:

Li et al.

Appl. No. 09/720,086

102(e): July 23, 2001

For:

De Novo DNA Cytosine
Methyltransferase Genes,
Polymentides and View The

Polypeptides and Uses Thereof

Confirmation No.: 6968

Art Unit:

1642

Examiner:

Harris, A. M.

Atty. Docket: 0609.4560002/KRM/DJN

Declaration Under 37 C.F.R. § 1.132 of Kenneth D. Bloch, M.D.

Commissioner for Patents P.O. Box 1450 Alexandria, VA 22313-1450

Sir:

I, the undersigned, Kenneth D. Bloch, M.D., residing at 80 Park Street, Apt. 32, Brookline, Massachusetts, 02446, declare and state as follows:

- 1. I am currently employed by Massachusetts General Hospital in the Cardiovascular Research Center where I conduct and supervise research in the field of molecular cardiology. I worked with Dr. Bn Li, co-inventor of the captioned application, in the Cardiovascular Research Center for 10 years until 2003 when Dr. Li moved to Novartis. I am also an Associate Professor of Anesthesiology and Medicine at Harvard Medical School and have extensive experience in molecular biology and DNA cloning and sequencing.
 - A current curriculum vitae is appended hereto as EXHIBIT A.
- 3. I have been informed that human DNMT3A cDNA clone, represented in the captioned patent application as SEQ ID NO:3, was deposited with the ATCC on July 10, 1998, and assigned ATCC Deposit No. 98809. I have also been informed that the deposit date of July 10, 1998 was prior to the filing date of the second provisional application, App. No. 60/093,993, filed July 24, 1998, the benefit of which is claimed. Finally, I have been informed that the '993 application includes the sequence information and references the deposit of the sequenced material on page 16, lines 1-2, of the specification.
- 4. In November 2004, Applicants had samples withdrawn of the human DNMT3A cDNA clone contained within ATCC Deposit No. 98809. At the Applicants request, I have sequenced the nucleotides that span the coding region of the deposited human DNMT3A cDNA clone contained in ATCC Deposit No. 98809. A nucleotide alignment that spans the coding regions of the sequenced human DNMT3A cDNA clone contained in

ATCC Deposit No. 98809 and currently amended SEQ ID NO:3 is shown in Fig. 1 of EXHIBIT B.

- 5. The amendment to the sequence listing, which was filed on July 23, 2001, corrected six nucleotide errors in the coding sequence of SEQ ID NO:3 (see bolded nucleotides at nucleotide positions 940, 1476, 1479, 1570, 2024 and 2119 of amended SEQ ID NO:3 in Fig. 1 of EXHIBIT B). The amendment also deleted original nucleotides 1-123 of SEQ ID NO:3, which does not include any DNMT3A coding sequence.
- 6. The deposited clone recited in ¶4 and 5, above (i.e., ATCC Deposit No. 98809) is the same as the deposited clone recited in the above-captioned application. The six nucleotides in the coding region of SEQ ID NO:3 that were corrected by the amendment of July 23, 2001 correspond to the sequence contained in ATCC Deposit No. 98809. It is well known that sequencing errors are a common problem in Molecular Biology. Peter Richterich, Estimation of Errors in Raw' DNA Sequences: A Validation Study, 8 Genome Research 251-59 (1998)(EXHIBIT C). I believe that one skilled in the art would have sequenced the deposited material and recognized the sequencing errors.
- 7. My sequencing of ATCC Deposit No. 98809 also revealed that nucleotides 539-584 within the coding region of amended SEQ ID NO:3 are deleted in the deposited cDNA. The deletion causes a frame shift in the coding region of the deposited cDNA and predicts a truncated protein of 145 amino acids. An amino acid alignment of the predicted amino acid sequence encoded by the human DNMT3A cDNA clone contained in ATCC Deposit No. 98809 and the predicted amino acid sequence encoded by amended SEQ ID NO:3 is shown in Fig. 2 of EXHIBIT B. The bolded sequence in Fig. 2 corresponds to the predicted encoded amino acid sequence downstream of the nucleotide deletion in ATCC Deposit No.98809 and represents a point of divergence compared with the predicted amino acids encoded by currently amended SEQ ID NO:3.
- 8. Currently amended SEQ ID NO:3 and SEQ ID NO:3 as originally filed in U.S. Appl. Nos. 60/090,906 and 60/093,993, to which priority is claimed, do not harbor the deletion, and encode a protein having 912 amino acids that is homologous to mouse Dnmt3a.
- 9. Like DNA sequence errors, it is known that errors in DNA cloning may occur in molecular biology. Deletion errors may occur and may be caused by, inter alia, inadvertent digestion of DNA by restriction endonucleases or exonucleases, or by recombination events during propagation of the DNA in bacterial hosts. The deletion at nucleotides 539-584 in SEQ ID NO:3 present in ATCC Deposit No. 98809 is an obvious error. I believe that one skilled in the art would have sequenced the deposited material and recognized the deletion as an error. My belief is based upon the following: First, the deletion found in ATCC Deposit No. 98809 is not present in SEQ ID NO:3 as originally filed or as amended. Second, the deletion is not present in the DNA sequence of the closely related mouse homolog, SEQ ID NO:1. Third, the deletion causes a frame shift in the reading frame of SEQ ID NO:3 and predicts a truncated protein product compared with that

encoded by SEQ ID NO:3 as originally filed and as amended. Fourth, the amino acids encoded by the nucleotide sequence downstream of the deletion bear no similarity to the amino acids encoded by SEQ ID NO:3 or the mouse homolog of Dnmt3a, encoded by SEQ ID NO:1. Finally, an examination of the sequence reveals two large open reading frames (ORF) in the sequence in different frames. See Fig. 3 of EXHIBIT B. The ORFs correspond to the amino acid residues of DNMT3A upstream and downstream of the deletion. The presence of two large ORFs in different frames indicates a possible frameshifting sequence error or deletion. All of these factors indicate that the deletion present in ATCC Deposit No. 98809 is an error, and would be recognized as such by a person of ordinary skill in the art.

- 10. It is my belief that the combination of ATCC Deposit No. 98809, which discloses the six nucleotides in the coding region of SEQ ID NO:3 amended on July 23, 2001, in combination with SEQ ID NO:3 as originally filed in U.S. Appl. Nos. 60/090,906 and 60/093,993, which disclose nucleotides 539-584 of amended SEQ ID NO:3, clearly conveys to someone skilled in the art the entire nucleotide sequence of amended SEQ ID NO:3.
- 11. I hereby declare that all statements made herein of my own knowledge are true and that all statements made on information and belief are believed to be true; and further that these statements were made with the knowledge that willful false statements and the like so made are punishable by fine or imprisonment, or both, under Section 1001 of Title 18 of the United States Code and that such willful false statements may jeopardize the validity of the present patent application or any patent issued thereon.

Respectfully submitted,

Kenneth D. Bloch, M.D.

Date: 11 7 2005

EXHIBIT A

CURRICULUM VITAE

PART I: General Information

DATE PREPARED: July 5, 2005

Name: Kenneth D. Bloch

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Place of Birth: New York, NY

Education:

1978 Brown University (Biomedicine) Sc.B.

1981 M.D. Brown University

Postdoctoral Training:

Internship and Residencies:

1981-1984 Internal Medicine, Massachusetts General Hospital

Fellowships:

1981-1984 Clinical Fellow in Medicine, Harvard Medical School 1984-1987 Research Fellow in Genetics, Harvard Medical School 1987-1989

Clinical and Research Fellow in Medicine, Harvard Medical School

Licensures and Certification:

1984-Commonwealth of Massachusetts, Registered Physician, 1985

Diplomate, American Board of Internal Medicine 1989

Diplomate, Subspecialty Board of Cardiovascular Diseases,

American Board of Internal Medicine

Academic Appointments:

1989-1990 Instructor in Medicine, Harvard Medical School

1990-1997 Assistant Professor of Medicine, Harvard Medical School 1997- Associate Professor of Medicine, Harvard Medical School
Associate Professor of Anaesthesia, Harvard Medical School

Hospital or Affiliated Institution Appointments:

1990-1996	Assistant in Medicine, Massachusetts General Hospital
1996-1999	Assistant Physician, Massachusetts General Hospital
1999-2005	Associate Physician, Massachusetts General Hospital
2005-	Physician, Massachusetts General Hospital

Other Professional Positions and Major Visiting Appointments: none

Hospital and Health Care Organization Service Appointments:

1990- Attending Physician, Cardiology and Medical Services,
Massachusetts General Hospital
Practice Member, Cardiac Unit Associates,
Massachusetts General Hospital

Major Administrative Responsibilities:

1989-1992 1990-	Principal Investigator, Cardiac Unit, Massachusetts General Hospital Associate Program Director, Fellowship Program in Cardiovascular
1992-	Disease, Massachusetts General Hospital Principal Investigator, Cardiovascular Research Center, Massachusetts General Hospital
1993-1997	
1997-2002	Preceptor, Training Grant to the Cardiovascular Research Center (PI: Mark C. Fishman), Massachusetts General Hospital Co-Investigator, Training Grant to the Cardiovascular Research Center (PI: Mark C. Fishman), Massachusette Grandiovascular Research Center
2002-2004	Control of a control of the second of the se
2002 2004	medim Director, Cardiovascular Research Center Manage
2002-	General Hospital Principal Investigator, Training Grant to the Cardiovascular Research Center, Massachusetts General Hospital

Major Committee Assignments:

Hospital: 1991	Member, Selection Committee for the Fellowship Program in Cardiovascular Disease, Massachusetts General Hospital
1992-1996 1997-	on Research, Massachusetts General Hospital
	Member, Steering Committee coordinating the integration of the clinical cardiology fellowship programs at Brigham and Women's Hospital and Massachusetts General Hospital

Regional: 1991-1994 Member, Research Peer Review Committee, American Heart Association, Massachusetts Affiliate, Inc. 2002 Member, Northeast Peer Review Study Group for Lipids, Thrombosis & Vascular Wall Biology, American Heart Association National: 1992-1995 Member, Molecular Signaling Research Study Committee, American Heart Association, National Center 1996-1997 Member, Lung Biology and Pathology Study Section, National Heart Lung and Blood Institute 2000 Member-At-Large, Executive Committee of the Council on Cardiopulmonary and Critical Care, American Heart Association, National Center 2000-2002 Member, Program Committee, Council on Cardiopulmonary and Critical Care, American Heart Association, National Center 2002-Chairman, Program Committee, Council on Cardiopulmonary and Critical Care, American Heart Association, National Center 2002-Member, Committee on Scientific Sessions Program, American Heart Association 2005 Member, Respiratory Integrative Biology and Translational Research Study Section, National Heart Lung and Blood Institute 2005-Member, Research Committee, American Heart Association 2005-Member, Future of Scientific Sessions Task Force, American Heart

Professional Societies:

American Heart Association, Council on Basic Cardiovascular Science
American Heart Association, Council on Cardiopulmonary and Critical
Care
American Society for Clinical Investigation

Editorial Boards:

1989- Ad hoc reviewer:

American Journal of Pathology American Journal of Physiology

American Journal of Respiratory Cell and Molecular Biology

Anesthesiology Circulation

Circulation Research

Journal of Applied Physiology Journal of Biological Chemistry Journal of Clinical Investigation New England Journal of Medicine

Nature Medicine

Proceedings of the National Academy of Sciences

Awards and Honors:

1978	Magna cum laude, Brown University
1981	Patricia McCormick Memorial Prize, Brown University
1984-1986	Research Fellowskin I and Service Stown University
1986-1987	Research Fellowship, Leukemia Society of America
1996-1998	Postdoctoral Fellowship, Pfizer Pharmaceuticals
	Grant-In-Aid, American Heart Association, National Courts of the
1996	Finalist, Circulation Council Cardiovascular Research Competition
2000	South Tize for the Dasic science recearch character
2004	from the Massachusetts Thoracic Society
2001	Charter Fellow of the Council on Basic Cardiovasculor Saint
0000	Amorican rieart Association
2002	Inaugural Fellow of the Basic Sciences Council, American Heart
	Association Association
2003	Inaugural Fellow of the Council on Cardiopulmonary, Perioperative, and Critical Care, American Heart Association
	1100001441011

Part II: Research, Teaching, and Clinical Contributions

A. Narrative Report

Dr. Bloch's research has focused on cardiovascular biology and the molecular mechanisms regulating vascular tone and ventricular remodeling. Dr. Bloch established his own laboratory at MGH in 1991 and became a principal investigator in the Cardiovascular Research Center in 1992.

When he established his laboratory, it appeared that there was a single constitutivelyexpressed nitric oxide (NO) synthase (NOS) in brain and endothelium. Dr. Bloch's group cloned the NOS isoform responsible for endothelial NO production, NOS3. Dr. Bloch's group discovered that NOS3-deficient mice are more susceptible to hypoxia-induced pulmonary vascular remodeling. Enhanced pulmonary NO (either by inhalation of NO gas or by aerosol delivery of an adenovirus specifying NOS attenuates pulmonary vasoconstriction and prevents this pulmonary vascular remodeling. Moreover, they found that NO inhalation could prevent pulmonary vascular remodeling even before the development of pulmonary vasoconstriction. Taken together, these studies have important implications for the treatment of children with congenital heart disease, in whom pulmonary vascular remodeling precedes the development of pulmonary hypertension (which is often fatal). Dr. Bloch and his colleagues were the first to show that NO inhalation also has systemic vascular effects including attenuation vascular neointima formation after balloon injury and prevention re-thrombosis after coronary artery thrombolysis (the latter is markedly potentiated by inhibitors of cGMPmetabolizing phosphodiesterases). These innovative studies have direct clinical implications for the care of patients with acute coronary syndromes.

Dr. Bloch has brought into focus the fact that regulation of NO <u>responsiveness</u> may be as important as regulation of NO <u>production</u>. Dr. Bloch found that prolonged exposure of vascular smooth muscle cells (VSMC) to NO or pro-inflammatory cytokines decreases function of soluble guanylate cyclase (sGC; the enzyme responsible for cGMP synthesis in response to NO), desensitizing the cells to NO. Dr. Bloch also showed that cGMP-dependent protein kinase, an enzyme responsible for vasodilation, also has a critical role in determining the sensitivity of these cells to the antiproliferative and proapoptotic effects of NO and cGMP.

Dr. Bloch's group has contributed importantly to a second research area--the structure and function of the "nuclear body", a nuclear structure that appears to have a critical role in oncogenesis, gene transcription, and the cellular response to viral infection. They have cloned two new nuclear body components, both of which appear to be transcription factors and one of which is a co-activator of nuclear hormone receptors.

Dr. Bloch's research group has elucidated an important role for NO synthesized by endothelial nitric oxide synthase (NOS3) in the left ventricular remodeling induced by a variety of hemodynamic challenges. In addition, they have explored the mechanisms responsible for the impairment of hypoxic pulmonary vasoconstriction associated with pulmonary injury associated with endotoxemia and volutrauma.

Most recently, Dr. Bloch's group has begun a new line of research directed at understanding how mutations in the gene encoding bone morphogenetic protein receptor type 2 (BMPR2) cause primary pulmonary hypertension. They have observed that mice carrying one copy of a mutant BMPR2 gene have mild pulmonary hypertension associated with abnormalities of pulmonary vascular structure.

Dr. Bloch has also made important contributions in the clinical and research training of cardiology fellows. He is a primary research mentor for the MGH cardiology fellows guiding them to, and supporting them in, the best research training opportunities HMS has to offer. In the past five years, Dr. Bloch has played a pivotal role in the integration of the cardiology fellowship programs at the Brigham and Women's Hospital and MGH providing fellows with exposure to outstanding clinical experiences at both institutions. Since 2002, Dr. Bloch has served as principal investigator of the T32 training grant awarded to the Cardiovascular Research Center. In this role, Dr. Bloch supervises the mentoring and career development of 10 post-doctoral cardiovascular scientists each year. From 2002 through 2004, Dr. Bloch served as the Interim Director of the CVRC fostering the creativity and productivity of ten faculty members and more than 40 post-doctoral research fellows.

B. Funding Information (Research):

Past:	
1991-1992	Research Grant, Dr. Louis Skarow Memorial Fund (PI: K. Bloch)
1991-1994	"Gene expression in a model of pulmonary hypertension." Grant-in-Aid, American Heart Association, National Center (PI: K. Bloch) "Pulmonary expression of endothelin genes."
,	O 1 1

1991-1996	NHLBI/R29 (PI: K. Bloch)
	"Biosynthesis of the endothelin family of vasoactive peptides."
1995-1996	Sponsored Research Agreement through the Cardiovascular
	Research Center from Printed M. G. His Cardiovascular
	Research Center from Bristol-Myers Squibb (PI: K. Bloch)
	"Isolation and characterization of novel vascular nitric oxide
1996-1997	syntases.
1990-199/	NHLBI/R01 (Co-PI: K. Bloch)
1006.000	"The pulmonary response to inhaled particulates."
1996-2000	NULDIKUI (PI: K. Bloch)
	"Nitric oxide/cGMP signal transduction in pulmonary injury."
1996-2001	Established Investigator, American Heart A.
	Established Investigator, American Heart Association, National Center
	(PI: K. Bloch) "Regulation of a nitric oxide receptor component, the β1
1998-2002	and with or politoic guallylate cyclage "
2002	NHLBI/T32 (Co-PI: K. Bloch; PI: M.C. Fishman)
1998-2003	"Cell and molecular training for cardiovascular biology."
1770-2003	TATEDURUI (PI: K. Bloch)
2001-2002	"Nitric oxide/cGMP signal transduction in vascular injury."
2001-2002	1 11201 1 Harmaceulicais (Co-Pi · K Bloch · Di · M · d · · · ·
	Evaluation of the effects of sildenafil with and with and the
	oxide (140) oil platelet-mediated thrombosis and conding front.
	canine model of cyclic coronary artery occlusion."
	y and obtaining unterly occidisfoll.
Current:	·
1996-	NHLBI/R01 (Co-PI: K. Bloch; PI: W. Zapol)
	"Studies of inhaled nitric oxide."
2002-	INO Therapeutics, Inc. (PI: K Bloch)
	"Laboratory based initiation C. of a
	"Laboratory-based initiatives for the further development of the
2002-	anotapoutic potential of innaled nitric oxide "
2002-	NHLBI/T32 (PI: K. Bloch)
2003-	"Cell and molecular training for cardiovascular biology."
2005-	NILDIKUI (PI: K. Bloch)
2003-	"Nitric oxide synthase 3 and left ventricular remodeling"
ZUU3-	NHLBI/R01 (PI: K. Bloch)

C. Report of Current Research Activities (Bench and Clinical Research):

Project 1: Studies of inhaled nitric oxide. (Co-PI: K. Bloch)

NHLBI/R01 (PI: K. Bloch)

Project 2: Evaluation of the systemic effects of breathing nitric oxide (PI: K. Bloch)

Project 3: Role of nitric oxide in left ventricular remodeling (PI: K. Bloch)

Project 4: Role of BMPR2 in the pathogenesis of pulmonary hypertension (PI: K. Bloch) Project 5: Evaluation of nitric oxide inhalation to treat cardiogenic shock complicating right ventricular infarction. (Co-PI: K. Bloch)

"BMPR2 in the pathogenesis of pulmonary hypertension"

D. Report of Teaching

1. Local Contributions

a. medical school

1977

Brown University, Biomed 110, Biophysics

course director: Babette Stewart

Teaching Assistant

50 undergraduates (approx.)

3-5 hours preparation and contact time/week (approx.)

semester course

1978-1981

Brown University, Biomed 6, Introduction to Physiology

course director: Peter Stewart

Teaching Assistant

50 undergraduates (approx.)

3-5 hours preparation and contact time/week (approx.)

semester course

1986

Harvard University, Genetics 700.0, Fundmentals of Genetics

course director: Philip Leder

Teaching Assistant

15 medical students (approx.)

3 hours preparation and contact time/week (approx.)

semester course

1990-1992

Harvard Medical School, Introduction to Clinical Medicine

Clinical Mentor

2-3 medical students, 3 sessions/week, 2 hours/session, 3 weeks/year

total: 9 hours preparation time, 18 hours contact time

1993-1996

Harvard Medical School, Patient-Doctor II Course

course directors: Katherine Treadway and Diane Fingold

Preceptor

50 medical students (approx.) 3 sessions, 2 hours/session, 1/year

total: 9 hours preparation and contact time

b. graduate medical course: none

c. local invited teaching presentations

1993-1994

"Ethical Conduct of Research" Course, Massachusetts General

Hospital, Preceptor, 5-10 postdoctoral fellows, 2 sessions/year,

2 hours/session, prep time: 1 hour/session.

1994

Cardiology Grand Rounds, Massachusetts General Hospital, Lecturer;

1996	50 attendees: medical students, residents, clinical and research fellows, faculty; 4 hrs prep and 1 hr contact time. Anesthesiology Grand Rounds, Massachusetts General Hospital;
1995	research fellows, faculty; 4 hrs prep and 1 hr contact time. Wellman Laboratories of Photomedicine Symposium, Massachusetts General Hospital, Session Chair: 50 attendes and the contact time.
1997	Cardiology Grand Rounds, Massachusetts General Hospital, Lecturer;
1998	Clinical Fellows' Core Curriculum Lacture G
1999	4 hrs prep time and 1 hr contact time. Pulmonary and Critical Care Unit Personal Care Contact time.
2000	clinical and research fellows, faculty; 4 hrs prep and 1 hr contact time. Center for the Prevention of Cardiovascular Disease, Department of Nutrition, Harvard School of Public Health I.
2001	4 hrs prep and 1 hr contact time. West Roxbury Veterans Administration Hospital, Cardiology Grand Rounds; 30 attendees: medical students, residents, residents.
2003	Brigham and Women's Hospital Vaganter P
2004	faculty; 4 hrs prep and 2 hr contact time. Massachusetts General Hospital, Cardiology Grand Rounds; 50 attendees: nurses, medical students, residents, elicitations.
2005	fellows, faculty; 5 hrs prep and 1 hr contact time. Massachusetts General Hospital, Critical Care Research Retreat; 50 attendees: clinical and research fellows, faculty; 10 hours prep time, 4 hours contact time, 10 minute lecture
d continuing	

d. continuing medical education courses: none

e. advisory and supervisory responsibilities

1989-	Attending Physician, Private and Ward Medical Services,
1989-1992	Principal Investigator, Cardiovascular Research Center, Massachusetts General Hospital scientific supervises for B
1990-	(1,000 hours/year). Attending Physician, Cardiology Consult Somice M.
1991-	General Hospital (140 hours/year). Research Advisor, Cardiology Fellowship Program, Massachusetts

1992-2002	General Hospital, 25-35 Fellows per year in clinical and research training (200 hours/year).
	Attending Physician, Coronary Care Unit, Massachusetts General Hospital (variable hours/year).
1992-	Principal Investigator, Cardiovascular Research Center, Massachusetts General Hospital, scientific supervisor for Personal P. II
	including two Assistant Professors of Anesthesia (1,000 hours/year). Principal Investigator, NIH-sponsored program (T32) to train 10 post-doctoral cardiovascular scientists each year (200 hours/year)

f. teaching leadership role

1990-1999	Cardiac Unit Research Seminar Series, Massachusetts General Hospital; Organizer; A weekly series of presentation by staff and
1995-1996	senior fellows designed to highlight research in the Cardiac Unit. Cardiac Unit Society of Fellows, Massachusetts General Hospital; Organizer; A quarterly series of symposia presented by research
1997	Society of Cardiology Fellows, Massachusetts General Hospital and Brigham and Women's Hospital; Co-Organizer; A quarterly series of symposia designed to foster scientific communication.
2002-	collaboration between MGH and BWH and to highlight research opportunities for fellows in cardiology training. CVRC Seminar Series; Organizer; A weekly seminar series presented by visiting scientists in the MGH Cardiovascular Research Center

g. names of advisees and trainees/current positions

1989-1990	Charles C. Hong, MD/PhD, Clinical and Research Fellow,
1990-1991 1991-1993	Robert Schott, MD, MPH, private practice
1990-1991	Akito Shimouchi, MD, Assistant Professor of Medicine, National Cardiovascular Center Research Institute, Osaka, Japan Stefan P. Janssens, MD, PhD, Associate Professor of Medicine, Cardiac Unit and Laborator of Medicine,
1992-1994, 2001-	University Hospital Gasthuisberg, Leuven, Belgium Noriko Kawai, MD, PhD, Research Fellow in Modician
1992-1993	John J. Lepore, MD. Instructor of Medicine, University of
	Cardiovascular Medicine Division and Molecular Cordials
1993-1994	Johanna Wolfram, MD. University Clinic for Internal No. 11.
1993-1999	Department of Cardiology, General Hospital, Vienna, Austria Lucienne Sanchez, MD, Instructor in Pediatrics, Harvard Medical School/Massachusetts General Hospital

1993-2004	Jesse D. Roberts, MD, Associate Professor of Anesthesiology in
1004 1005	i culaules, marvard Medical School/Massachusetta Gomena III. audical
1994-1995	Joiney Hollias, MD. Associate Professor of Neurosumann. Cl.: C. c.
	riculovascular and NeuroInterventional Surgery Division of
	Neurosurgery, The University of New Mexico Health Sciences
	Center
1994-1996	Alexandra Holzmann, MD, Department of Anesthesiology,
	University of Heidelberg
1995-2004	Heling Liu, MD, on leave to care for her children
1995-1998	Jean-Daniel Chiche, MD, Professor, Cochin University, Paris,
	France France
1996-1997	Anita Honkanen, MD, private practice
1996-1998	Masao Takata, MD/PhD, Assistant Professor, Department of
	Anaesthetics and Intensive Care, Imperial College School of
	Medicine, Hammersmith Hospital, London, United Kingdom
1996-1998	Douglas Wirthlin, MD, Assistant Professor of Surgery, Vascular
	Surgery, University of Alabama at Birmingham
1996-1998	Joerg Weimann, MD, Professor of Anesthesia, Department of
	Anaesthesiology and Intensive Care Medicine, Charité -Berlin
	Medical School, Campus Benjamin Franklin
1998-2002	Galina Filippov, MD, Research Scientist, Omnigene Bioproducts
	Inc.
1998, 2000-2001	Zena Quezado, MD, Chief, Department of Anesthesia and Surgical
	Services, Warren G. Magnuson Clinical Center, National Institutes
	of Health
1998-2000	Roman Ullrich, MD, Associate Professor of Anesthesia and
	Intensive Care Medicine, Vienna General Hospital, Medical
	University of Vienna
1999-2001	Hiroshi Nakajima, MD, Neurosurgery Residency, Tokyo Women's
	Medical College, Tokyo, Japan
1999-2001	Pini Orbach, PhD, Project Manager, Drug Development, Perdix
	Pharmaceuticals, Inc.
1999-	Fumito Ichinose, MD, PhD, Assistant Professor of Anesthesia,
	narvard Medical School/Massachusetts General Hognital
2001-2003	Aimee Limbach, PhD, Post-doctoral Fellow, Center for Human
	Molecular Genetics, Munroe-Meyer Institute, University of Nebraska
	Medical Center
2002-2003	Elisabeth Choe, MD, Resident in Internal Medicine, University of
	Texas Southwestern
2002-2004	Cornelius Busch, MD, Resident in Anesthesia, Department of
	Anesthesiology, University of Heidelberg
2003	Claire Mayeur, MD, Resident in Anesthesiology 1:110 France
2003-	Hideyuki Beppu, MD, PhD, Instructor in Medicine, Cardiovascular
	Research Center, Massachusetts (teneral Hospital
2003-	Manu Buys, PhD, Research Fellow in Medicine, Cardiovascular
	Research Center, Massachusetts General Hospital
	, Gonorai Hospital

2003-	Paul Yu, MD, PhD, Clinical and Research Fellow in Medicine,
2004-2005 2004-	David Bayne, undergraduate student, Harvard University Ryuji Hataishi, MD, PhD, Research Fellow in Aposthogic
2004-	wassachuseus General Hospital
2004-	Rajeev Malhotra, MD, Resident in Internal Medicine, Massachusetts General Hospital Tomas Neilan, MD, Clinical and Research Fellow in Medicine,
2005-	Cardiac Ultrasound Laboratory and Cardiovascular Research Center, Massachusetts General Hospital Sarah Blake, MD, Research Fellow in Anesthesia, Massachusetts General Hospital

2. Regional, National, or International Contributions

1994	Invited Lecturer American College CG 111
1994	Invited Lecturer, American College of Cardiology, Dallas, TX
1994	Invited Lecturer, Pfizer Pharmaceuticals, Groton, CT
1995	Invited Lecturer, University of Leuven, Belgium
	Invited Lecturer, St. Elizabeth's Hospital, Cardiovascular Research
1996	Sommar, Doston, IVIA
1996	Invited Lecturer, Boston Heart Foundation, Boston, MA
1790	mivited Lecturer, Georgia Medical College Vascular Dialams
1997	Division, Anama, GA
1997	Invited Lecturer, Harvard Medical School, Vascular Biology
100=	benimar, boston, MA
1997	Invited Lecturer, Boston University Whittaker Foundation B
1998	Invited Lecturer, Oregon Health Sciences University, Cardiology
	Division, Oregon
1998	Invited Lecturer, Brigham and Women's Hospital, Cardiology
	Division, Monday Morning Research Conference, Boston, MA
1998	Invited Lecturer, New York Medical College, Department of
	Pharmacology Seminar, New York, NY
1999	Invited Lecturer Tuba II-ii- G. t
•	Invited Lecturer, Tufts University School of Medicine, Dept. of Medicine, Boston, MA
1999	Invited Lectures Tugati
	Invited Lecturer, Tufts University School of Medicine/New England
1999	
1999	mytica Doctuici, Willenium Pharmacouticala Inc. Court in the
	my red botture, University of Washington Cardiology Dans
1999	South, WA.
1799	Invited Lecturer, University of Alabama at Birmingham, Dept. of
1999	* WALCIUE V. DITTOTTOTTOTA A I
1999	Invited Lecturer, 3 rd International Society for Modical Garage
1000	ricomig, ricidelogig, Germany
1999	Invited Lecturer, National Institute of Health, Critical Care
2004	Wediene, Beinesda, MI)
2004	Invited Lecturer, INO Therapeutics Inc. Scientific Advisory Board,
	Chatham, MA
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2002	Invited Lecturer, Medical College of Wisconsin, Milwaukee,
2002	Invited Lecturer, American Heart Association Scientific Co.
2003	Invited Lecturer, Department for Moleculer and B:
2003	Research, Universiteit Gent, Belgium Invited Lecturer, Whitaker Cardiovascular Institute, Boston University Medical School
2003	University Medical School Invited Lecturer, Cardiology Division, University of Alberta, Edmonton, Canada
2004	Invited Lecturer, American Heart Association
2004	
2004	Invited Lecturer, Cardiology Grand Rounds, Dartmouth-Hitchcock Medical Center—"Nitric oxide synthases in ventricular remodeling: insights gained from genetically-modified mice." Invited Lecturer, Vascular Biology Seminar, Dartmouth-Hitchcock Medical Center—"Mechanisms regulating pulmonary vascular structure and function—roles of land.
2004	morphogenetic proteins." Invited Lecturer, Department of Physiology Seminar, Louisiana State University-Shreveport—"Nitric oxide graph.
2004	Invited Lecturer, Cardiovascular Colland G.
2004	II, Cambridge, MA—"Nitric oxide/cGMP signal transduction—implications for cardiovascular gene transfer." Invited Lecturer, Critical Therapeutics, Inc., Lexington, MA—"Mechanisms of pulmonary vascular dysfunction in lung injury:insights gained from genetically-modified mice."

E. Report of Clinical Activities

Dr. Bloch is a practicing cardiologist who maintains a practice within the Cardiac Unit Associates and Cardiology Division at the Massachusetts General Hospital. His practice consists of patients with cardiac problems of a moderate to high level of complexity referred to a tertiary care center.

Part III: Bibliography

Original Articles:

- 1. Seidman CE, Bloch KD, Klein KA, Smith JA, Seidman JG. Nucleotide sequences of the human and mouse atrial natriuretic factor genes. Science 1984, 226:1206-1209.
- 2. Bloch KD, Scott JA, Zisfein JB, Fallon JT, Seidman CE, Matsueda GR, Margolies MN, Homcy CJ, Graham RM, Seidman JG. Biosynthesis and secretion of proatrial natriuretic factor by cultured rat cardiocytes. Science 1985, 230:1168-1172.
- 3. Graham RM, Bloch KD, Delaney VB, Bourke E, Seidman JG. Bartter's syndrome and the atrial natriuretic factor gene. Hypertension 1986, 8:549-551.
- 4. Ballerman BJ, Bloch KD, Seidman JG, Brenner BM. Atrial natriuretic peptide transcription, secretion, and glomerular receptor activity during mineralocorticoid escape. J Clin Invest 1986, 78:840-843.
- 5. Zisfein JB, Matsueda GR, Fallon JT, Bloch KD, Seidman CE, Seidman JG, Homcy CJ, Graham RM. Atrial natriuretic factor: assessment of its structure in atria and regulation of its biosynthesis with volume depletion. J Mol Cell Cardiol 1986, 18:917-929.
- 6. Bloch KD, Seidman JG, Naftilan JD, Fallon JT, Seidman CE. Neonatal atria and ventricles secrete atrial natriuretic factor via tissue-specific secretory pathways. Cell 1986, 47:695-702.
- 7. Bloch KD, Zisfein JB, Margolies MN, Homcy CJ, Seidman JG, Graham RM. Atrial natriuretic factor biosynthesis: a serum protease cleaves proANF into a 14-kilodalton peptide and ANF. Am J Physiol 1987, 252:E147-E151.
- 8. Bloch KD, Jones SW, Preibisch G, Seipke G, Seidman CE, Seidman JG. Proatrial natriuretic factor is phosphorylated by rat cardiocytes in culture. J Biol Chem 1987, 262:9956-9961.
- 9. Zeller R, Bloch KD, Williams BS, Arceci RJ, Seidman CE. Localized expression of the atrial natriuretic factor gene during cardiac embryogenesis. Genes and Development 1987, 1:693-698.
- 10. Mendez RE, Pfeffer JM, Ortola FV, Bloch KD, Anderson S, Seidman JG, Brenner BM. Atrial natriuretic peptide transcription, storage, and release in rats with myocardial infarction. Am J Physiol 1987, 253:H1449-H1455.
- 11. Lee RT, Bloch KD, Pfeffer JM, Pfeffer MA, Neer EJ, Seidman CE. Atrial natriuretic factor gene expression in ventricles of rats with spontaneous biventricular hypertrophy. J Clin Invest 1988, 81:431-434.

- 12. Bloch DB, Bloch KD, Iannuzzi M, Collins FS, Neer EJ, Seidman JG, Morton CC. The gene for the alpha(i)1 subunit of human G protein maps near the cystic fibrosis locus. Am J Hum Gen 1988, 42:884-888.
- 13. Seidman CE, Wong D, Bloch KD, Seidman JG. Cis-acting sequences that modulate atrial natriuretic factor gene expression. Proc Natl Acad Sci USA 1988, 85:4104-4108.
- 14. Kim S, Ang S-L, Bloch DB, Bloch KD, Kawahara Y, Tolman C, Lee R, Seidman JG, Neer EJ. Identification of cDNA encoding a new alpha subunit of a human GTP-binding protein: expression of three alpha(i) subtypes in human tissues and cell lines. Proc Natl Acad Sci USA 1988, 85:4153-4157.
- 15. Bloch KD, Zamir N, Seidman CE, Seidman JG. Ouabain induces secretion of proatrial natriuretic factor by neonatal rat atrial cardiocytes. Am J Physiol 1988, 255:E383-E387.
- 16. Ladenson PW, Bloch KD, Seidman JG. Modulation of atrial natriuretic factor (ANF) by thyroid hormone: mRNA and peptide levels in hypothyroid, euthyroid, and hyperthyroid rat atria and ventricles. Endocrinology 1988, 123:652-657.
- 17. Lee RT, Brock TA, Tolman C, Bloch KD, Seidman JG, Neer EJ. Subtype-specific increase in G-protein α-subunit mRNA by interleukin 1β. FEBS Letters 1989, 2:139-142.
- 18. Bloch KD, Friederich SP, Lee M-L, Eddy RL, Shows TB, Quertermous T. Structural organization and chromosomal assignment of the gene encoding endothelin. J Biol Chem 1989, 264:10851-10857.
- 19. Bloch KD, Eddy RL, Shows TB, Quertermous T. cDNA cloning and chromosomal assignment of the gene encoding endothelin 3. J Biol Chem 1989, 264:18156-18161.
- 20. Lee M-L, Bloch KD, Clifford JA, Quertermous T. Functional analysis of the endothelin-1 gene promoter: evidence for an endothelial cell-specific cis-acting sequence. J Biol Chem 1990, 265:10446-10450.
- 21. Lee M-L, de la Monte S, Ng S-C, Bloch KD, Quertermous T. Expression of the potent vasoconstrictor endothelin in the central nervous system. J Clin Invest 1990, 86:141-147.
- 22. Bloch KD, Hong CC, Eddy RL, Shows TB, and Quertermous T. cDNA cloning and chromosomal assignment of the endothelin 2 gene: vasoactive intestinal contractor peptide is rat endothelin 2. Genomics 1991, 10:236-242.
- 23. Janssens SP, Shimouchi A, Quertermous T, Bloch DB, and Bloch KD. Cloning and expression of a cDNA encoding human endothelium-derived relaxing factor/nitric oxide synthase. J Biol Chem 1992, 267:14519-14522.

- 24. Roberts JD, Chen T-Y, Kawai N, Wain J, Dupuy P, Shimouchi A, Bloch KD, Polaner D, Zapol WM. Inhaled nitric oxide reverses pulmonary vasoconstriction in the hypoxic and acidotic newborn lamb. Circ Res 1993, 72:246-254.
- 25. Shimouchi A, Janssens SP, Bloch DB, Zapol WM, Bloch KD. Cyclic AMP regulates soluble guanylate cyclase β1 subunit gene expression in RFL-6 rat fetal lung fibroblasts. Am J Physiol 1993, 265:L456-L461.
- 26. Rovira I, Chen T-Y, Winkler M, Kawai N, Bloch KD, Zapol WM. Effects of inhaled nitric oxide on pulmonary hemodynamics and gas exchange in an ovine model of ARDS. J Appl Physiol 1994, 76:345-355.
- 27. Suen HC, Bloch KD, Donahoe PK. Antenatal glucocorticoid treatment corrects the pulmonary immaturity of congenital diaphragmatic hernia. Pediatr Res 1994, 35:523-529.
- 28. Cicila GT, Rapp JP, Bloch KD, Kurtz TW, Pravenec M, Kren V, Hong CC, Quertermous T, Ng S-C. Cosegregation of the endothelin-3, but not the endothelin-1, locus with blood pressure and relative heart weight in inbred Dahl rats. J Hypertension 1994, 12:643-651.
- 29. Horwitz MJ, Bloch KD, Kim NB, Amico JA. Expression of the endothelin 1 and oxytocin genes in the hypothalamus of the pregnant rat. Brain Research 1994, 648:59-64.
- 30. Bloch DB, Rabkina D, Quertermous T, Bloch KD. The immunoreactive region in a novel autoantigen contains a nuclear localization sequence. Clin Immunol Immunopath 1994, 72:380-389.
- 31. Roberts JD, Roberts CT, Jones RC, Zapol WM, Bloch KD. Nitric oxide inhalation reduces hypoxic pulmonary arterial remodeling, right ventricular hypertrophy, and growth retardation in the newborn rat. Circ Res 1995, 76:215-222.
- 32. de la Monte SM, Quertermous T, Hong CC, Bloch KD. Regional and maturation-associated expression of endothelin 2 in the gastrointestinal tract. J Histochem Cytochem 1995, 43:203-209.
- 33. Staples JF, Zapol WM, Bloch KD, Kawai N, Val VMF, Hochachka PW. Nitric oxide responses of air-breathing and water-breathing fish. Am J Physiol 1995, 268:R816-R819.
- 34. Kawai N, Bloch DB, Filippov G, Rapkina D, Suen H-C, Losty PD, Janssens SP, Zapol WM, de la Monte SM, Bloch KD. Constitutive endothelial nitric oxide synthase gene expression is regulated during lung development. Am J Physiol 1995, 268:L589-L595.

- 35. Bloch KD, Wolfram JR, Roberts JD, Zapol DG, Lepore JJ, Filippov G, Thomas JE, Brown D, Jacob HJ, Bloch DB. Three members of the nitric oxide synthase II gene family co-localize to human chromosome 17. Genomics 1995, 27:526-530.
- 36. Kurrek MM, Castillo L, Bloch KD, Tannenbaum SR, Zapol WM. Inhaled nitric oxide does not alter endotoxin-induced nitric oxide synthase activity during perfusion of the isolated rat lung. J Appl Physiol 1995, 79:1088-1092.
- 37. Bloch DB, Rabkina D, Bloch KD. The cell proliferation-associated antigen Ki-67 is a target of antibodies in the serum of MRL mice. Lab Invest 1995, 73:366-371.
- 38. Kurrek MM, Holzmann A, Filippov G, Winkler M, Zapol WM, Bloch KD. In vivo lipopolysaccharide pretreatment inhibits cGMP release from the isolated-perfused rat lung. Am J Physiol 1995, 269:L618-L624.
- 39. Huang PL, Huang ZH, Bloch KD, Moskowitz MA, Bevan JS, Fishman MC. Targeted disruption of the endothelial nitric oxide synthase gene causes hypertension. Nature 1995, 377:239-242.
- 40. Lee JS, Adrie C, Jacob HJ, Roberts JD, Zapol WM, Bloch KD. Chronic inhalation of nitric oxide inhibits neointima formation after balloon arterial injury in rats. Circ Res 1996, 78:337-342.
- 41. Janssens SP, Bloch KD, Nong Z, Gerard RD, Zoldhelyi P, Collen D. Adenovirus-mediated transfer of the human constitutive endothelial nitric oxide synthase gene to hypoxic rat lungs. J Clin Invest 1996, 98:317-324.
- 42. de la Monte SM, Bloch KD. Aberrant expression of the constitutive endothelial nitric oxide synthase gene in Alzheimer's disease. Mol Chem Neuropath 1997, 30:139-159.
- 43. Adrie C, Bloch KD, Moreno PR, Hurford WE, Guerrero L, Holt R, Zapol WM, Gold HK, Semigran MJ. Inhaled nitric oxide increases coronary artery patency after thrombolysis. Circulation 1996, 94:1919-1926.
- 44. Holzmann A, Bloch KD, Sanchez LS, Filippov G, Zapol WM. Hyporesponsiveness to inhaled NO in isolated-perfused lungs from endotoxin-challenged rats. Am J Physiol 1996, 271:L981-L986.
- 45. Bloch DB, de la Monte SM, Guigaouri P, Filippov A, Bloch KD. Identification and characterization of a leukocyte-specific component of the nuclear body. J Biol Chem 1996, 271:29198-29204.
- 46. Bloch KD, Filippov G, Sanchez LS, Nakane M, de la Monte SM. Pulmonary soluble guanylate cyclase, a nitric oxide receptor, is increased during the perinatal period. Am J Physiol 1997, 272:L400-L406.

- 47. Liu HL, Force T, Bloch KD. Nerve growth factor decreases soluble guanylate cyclase in rat pheochromocytoma PC12 cells. J Biol Chem 1997, 272:6038-6043.
- 48. Liu H-W, Anand A, Bloch KD, Christiani D, Kradin R. Expression of inducible nitric oxide synthase by macrophages in rat lung. Amer Rev Resp Dis 1997, 156:223-8.
- 49. Filippov G, Bloch DB, Bloch KD. Nitric oxide decreases stability of mRNAs encoding soluble guanylate cyclase subunits in rat pulmonary artery smooth muscle cells. J Clin Invest 1997, 100:942-948.
- 50. Head CA, Brugnara C, Martinez-Ruiz R, Kacmarek RM, Bridges KR, Kuter D, Bloch KD, Zapol WM. Low concentrations of nitric oxide increase oxygen affinity of sickle erythrocytes in vitro and in vivo. J. Clin. Invest. 1997, 100:1193-1198.
- 51. Sanchez LS, Filippov G, Zapol WM, Jones RC, Bloch KD. cGMP-binding, cGMP-specific phosphodiesterase gene expression is regulated during lung development. Pediatr. Res. 1998, 43:163-168.
- 52. Ichinose F, Adrie C, Hurford WE, Bloch KD, Zapol WM. Aerosolized zaprinast potentiates and prolongs the pulmonary vasodilation induced by breathing nitric oxide. Anaesthesiol. 1998, 88:410-416.
- 53. Steudel W, Scherrer-Crosbie M, Bloch KD, Weimann J, Huang PL, Picard MH, Zapol WM. Sustained pulmonary hypertension and right ventricular hypertrophy after chronic hypoxia in mice with congenital deficiency of nitric oxide synthase 3. J. Clin. Invest. 1998, 101:2468-2477.
- 54. Weimann J, Bauer H, Bigatello L, Bloch KD, Martin E, Zapol WM. ABO blood group and inhaled nitric oxide in acute respiratory distress syndrome. Lancet 1998, 351:1786-1787.
- 55. Chiche J-D, Schlutsmeyer SM, Bloch DB, de la Monte SM, Roberts JD, Filippov G, Janssens SP, Rosenzweig A, Bloch KD. Adenovirus-mediated gene transfer of cGMP-dependent protein kinase increases the sensitivity of cultured vascular smooth muscle cells to the antiproliferative and pro-apoptotic effects of nitric oxide/cGMP. J. Biol. Chem. 1998, 273:34263-34271.
- 56. Koike G, Chiche J-D, Shiozawa M, Simon JS, Szpirer J, Jacob HJ, Szpirer C, Bloch KD. Localization of rat genes in the nitric oxide signaling pathway: candidates for the pathogenesis of complex diseases. Mamm. Genome. 1999, 10:71-3.
- 57. Tyler RC, Muramatsu M, Abman SH, Stelzner TJ, Rodman DM, Bloch KD, McMurtry IF. Variable expression of endothelial NO synthase in three forms of rat pulmonary hypertension. Am. J. Physiol. 1999, 276:L297-303

- 58. Sohn YK, Ganju N, Bloch KD, Wands JR, de la Monte SM. Neuritic sprouting with aberrant expression of the nitric oxide synthase III gene in neurodegenerative diseases. J. Neurol. Sci. 1999, 162:133-51
- 59. Holzmann A, Manktelow C, Taut F, Bloch KD, Zapol WM. Inhibition of nitric oxide synthase prevents hyporesponsiveness to inhaled nitric oxide in lungs from endotoxin-challenged rats. Anesthesiology 1999, 91:215-21.
- 60. Powel V, Moreira GA, O'Donnell DC, Filippov G, Bloch KD, Gordon JB. Maturational changes in ovine pulmonary vascular responses to inhaled nitric oxide. Pediatric Pulmonology 1999, 27:157-66.
- 61. Bloch DB, Chiche J-D, Orth D, Rosenzweig A, Bloch KD. Structural and functional heterogeneity of nuclear bodies. Molecular and Cellular Biology 1999, 19:4423-4430.
- 62. Ullrich R, Bloch KD, Ichinose F, Steudel W, Zapol WM. Preservation of hypoxic pulmonary blood flow redistribution and arterial oxygenation in endotoxin-challenged mice with congenital NOS2 deficiency. J. Clin. Invest. 1999, 104:1421-1429.
- 63. Weimann J, Bloch KD, Takata M, Steudel W, Zapol WM. Congenital NOS2 deficiency protects mice from LPS-induced hyporesponsiveness to inhaled NO. Anesthesiology 1999, 91:1744-1753.
- 64. Weimann J, Ullrich R, Hromi J, Fujino Y, Clark M, Bloch KD, Zapol WM. Sildenafil is a pulmonary vasodilator in awake lambs with acute pulmonary hypertension. Anesthesiology 2000, 92:1702-1712.
- 65. Roberts JD, Chiche J-D, Weimann J, Steudel W, Zapol WM, Bloch KD. Nitric oxide inhalation decreases pulmonary artery remodeling in the injured lungs of rat pups. Circ. Res. 2000, 87:140-145.
- 66. Bloch DB, Nakajima A, Gulick T, Chiche J-D, Orth D, de la Monte SM, Bloch KD. Sp110 localizes to the PML/Sp100 nuclear body and may function as a nuclear hormone receptor transcriptional co-activator. Molecular and Cellular Biology 2000, 20:6138-6146.
- 67. Ullrich R, Scherrer-Crosbie M, Bloch KD, Ichinose F, Nakajima H, Picard MH, Zapol WM, Quezado ZMN. Congenital deficiency of nitric oxide synthase 2 protects against endotoxin-induced myocardial dysfunction in mice. Circulation 2000, 102:1440-1446.
- 68. Budts W, Pokreisz P, Nong Z, Van Pelt N, Gillijns H, Gerard R, Lyons R, Collen D, Bloch KD, Janssens S. Aerosol gene transfer with inducible nitric oxide synthase reduces hypoxic pulmonary hypertension and pulmonary vascular remodeling in rats. Circulation 2000, 102:2880-2885.

- 69. Sinnaeve P, Chiche J-D, Nong Z, Varenne O, Van Pelt N, Gillijns H, Collen D, Bloch KD, Janssens S. Soluble guanylate cyclase $\alpha 1$ and $\beta 1$ gene transfer increases NO responsiveness and reduces neointima formation after balloon injury in rats via antiproliferative and antimigratory effects. Circ. Res. 2001, 88:103-109.
- 70. Takata M, Filippov G, Liu H, Ichinose F, Janssens S, Bloch DB, Bloch KD. Cytokines decrease soluble guanylate cyclase in pulmonary artery smooth muscle cells via NO-dependent and NO-independent mechanisms. Am. J. Physiol. 2001, 280:L272-L278.
- 71. Holzmann A, Manktelow C, Weimann J, Bloch KD, Zapol WM. Inhibition of lung phosphodiesterase improves responsiveness to inhaled nitric oxide in isolated-perfused lungs from rats challenged with endotoxin. Intensive Care Medicine 2001, 27:251-257.
- 72. Ichinose F, Zapol WM, Sapirstein A, Ullrich R, Tager AM, Coggins K, Jones R, Bloch KD. Attenuation of hypoxic pulmonary vasoconstriction by endotoxemia requires 5-lipoxygenase in mice. Circ. Res. 2001; 88:832-838.
- 73. Ichinose F, Erana-Garcia J, Hromi J, Raveh Y, Jones R, Krim L, Clark MWH, Winkler JD, Bloch KD, Zapol WM. Nebulized sildenafil is a selective pulmonary vasodilator in lambs with acute pulmonary hypertension. Critical Care Medicine 2001, 29:1000-1005.
- 74. Schmidt U, Han RO, DiSalvo TG, Guerrero JL, Gold HK, Zapol WM, Bloch KD, Semigran MJ. Cessation of platelet-mediated cyclic canine coronary occlusion after thrombolysis by combining nitric oxide inhalation with phosphodiesterase inhibition. Journal of the American College of Cardiology 2001, 37:1981-1988.
- 75. Scherrer-Crosbie M, Ullrich R, Bloch KD, Nakajima H, Aretz HT, Lindsey ML, Vançon A-C, Nasseri B, Huang PL, Lee RT, Zapol WM, Picard MH. Nitric oxide synthase 3 limits left ventricular remodeling after myocardial infarction in mice. Circulation 2001; 104:1286-1291.
- 76. Raveh Y, Ichinose F, Orbach P, Bloch KD, Zapol WM. Radical scavengers protect murine lung from endotoxin-induced hyporesponsiveness to inhaled nitric oxide. Anesthesiology 2002, 96:926-33.
- 77. Ichinose F, Ullrich R, Sapirstein A, Jones RC, Bonventre JV, Serhan CM, Bloch KD, Zapol WM. Cytosolic phospholipase A₂ in hypoxic pulmonary vasoconstriction. J. Clin. Invest. 2002 109:1493-500.
- 78. Sinnaeve P, Chiche J-D, Gillijns H, Van Pelt N, Wirthlin DJ, Van de Werf F, Collen D, Bloch KD, Janssens S. Overexpression of a constitutively-active protein kinase G mutant reduces neointima formation and in-stent restenosis. Circulation 2002, 105:2911-6.

- 79. Lepore J, Maroo A, Pereira N, Ginns L, Dec G, Zapol W, Bloch K, Semigran M. Effect of sildenafil on the acute pulmonary vasodilator response to inhaled nitric oxide in adults with primary pulmonary hypertension. Am J Cardiol 2002, 90:677-680.
- 80. Baboolal HA, Ichinose F, Ullrich R, Kawai N, Bloch KD, Zapol WM. Reactive oxygen species scavengers prevent endotoxin-induced impairment of hypoxic pulmonary vasoconstriction in mice. Anesthesiology 2002, 97:1227-33.
- 81. de la Monte SM, Chiche J-D, von dem Bussche A, Sanyal S, Lahousse SA, Janssens SP, Bloch KD. Nitric oxide synthase-3 over-expression causes apoptosis and impairs neuronal mitochondrial function: relevance to Alzheimer-type neurodegeneration. Lab. Invest. 2003, 83:287-98.
- 82. Liu H, Nowak R, Chao W, Bloch KD. Nerve growth factor induces anti-apoptotic heme oxygenase-1 in rat pheochromocytoma PC12 Cells. J. Neurochem. 2003, 86:1553-63.
- 83. Wu JC, Nasseri BA, Bloch KD, Picard MH, Scherrer-Crosbie M. Influence of sex on ventricular remodeling after myocardial infarction in mice. J Am Soc Echocardiogr. 2003, 16:1158-62.
- 84. Ichinose F, Hataishi R, Wu JC, Kawai N, Tude-Rodrigues AC, Mallari C, Post JM, Parkinson JF, Picard MH, Bloch KD, Zapol WM. A selective inducible nitric oxide synthase dimerization inhibitor prevents systemic, cardiac and pulmonary hemodynamic dysfunction in endotoxemic mice. Am J Physiol 2003, 285:H2524-30.
- 85. Yu JH, Nakajima A, Nakajima H, Diller LR, Bloch KD, Bloch DB. Restoration of PML-nuclear bodies in neuroblastoma cells enhances retinoic acid responsiveness. Cancer Res 2004 64:928-33.
- 86. Ichinose F, Bloch KD, Wu JC, Hataishi R, Aretz HT, Picard MH, Scherrer-Crosbie M. Pressure-overload induced left ventricular hypertrophy and dysfunction in mice are exacerbated by congenital nitric oxide synthase 3 deficiency. Am J Physiol 2004 286:H1070-5.
- 87. Hassoun PM, Filippov G, Fogel, M, Donaldson C, Kayyali US, Shimoda LA, Bloch KD. Hypoxia decreases expression of soluble guanylate cyclase in cultures rat pulmonary artery smooth muscle cells. Am J Respir Cell Mol Biol. 2004, 30:908-13.
- 88. Janssens S, Pokreisz P, Schoonjans L, Pellens M, Vermeersch P, Tjwa M, Jans P, Scherrer-Crosbie M, Picard MH, Szelis Z, Gillijns H, Van de Werf F, Collen D, Bloch KD. Cardiomyocyte-specific over-expression of nitric oxide synthase 3 improves left ventricular performance and reduces compensatory hypertrophy after myocardial infarction. Circ Res, 2004, 94:1256-62.

- 89. del Monte F, Dalal R, Tabchy A, Couget J, Bloch KD, Peterson R, Hajjar RJ. Transcriptional changes following restoration of SERCA2a levels in failing rat hearts. FASEB J. 2004 18:1474-6.
- 90. Inglessis I, Shin JT, Lepore JJ, Palacios IF, Zapol WM, Bloch KD, Semigran MJ. Hemodynamic effects of inhaled nitric oxide in right ventricular myocardial infarction and cardiogenic shock. Journal of the American College of Cardiology 2004 44:793-8.
- 91. Tudes-Rodrigues AC, Hataishi R, Ichinose F, Bloch KD, Derumeaux G, Picard MH, Scherrer-Crosbie M. Relationship of systolic dysfunction to area at risk and infarction size after ischemia-reperfusion in mice. J Am Soc Echocardiogr 2004, 17:948-53.
- 92. Lewis G, Bloch KD, Semigran MJ. Pulmonary thromboembolism superimposed on a congenital VSD in a 50 year old man; inhaled nitric oxide and sildenafil to the rescue. Cardiology in Review 2004, 12:188-90.
- 93. Beppu H, Ichinose F, Kawai N, Jones RC, Yu P, Zapol WM, Miyazono K, Li E, Bloch KD. BMPR-II heterozygous mice have mild pulmonary hypertension and an impaired pulmonary vascular remodeling response to prolonged hypoxia. Am J Physiol 2004, 287:L1241-7.
- 94. Browner NC, Dey NB, Bloch KD, Lincoln TM. Regulation of cGMP-dependent protein kinase expression by soluble guanylyl cyclase in vascular smooth muscle cells. J Biol Chem. 2004, 279:46631-6.
- 95. Evgenov OV, Ichinose F, Evgenov NV, Gnoth MJ, Falkowski GE, Chang Y, Bloch KD, Zapol WM. Soluble guanylate cyclase activator reverses acute pulmonary hypertension and augments the pulmonary vasodilator response to inhaled nitric oxide in awake lambs. Circulation 2004, 110:2253-9.
- 96. Bloch DB, Yu JH, Yang W-H, Graeme-Cook F, Lindor K, Viswanathan A, Bloch KD, Nakajima A. The cytoplasmic dot staining pattern is detected in a subgroup of patients with primary biliary cirrhosis. Journal of Rheumatology, 2005, 32:477-83.
- 97. Weinberg EO, Scherrer-Crosbie M, Picard MH, Nasseri BA, MacGillivray C, Gannon J, Lian Q, Bloch KD, Lee RT. Rosuvastatin reduces experimental left ventricular infarct size following ischemia-reperfusion injury but not total coronary occlusion. Am J Physiol Heart Circ Physiol. 2005, 288:H1802-9.
- 98. Beppu H, Lei H, Bloch KD, Li E. Generation of a floxed allele of the mouse BMP type II receptor gene. Genesis 2005, 41:133-7.
- 99. Lepore JJ, Maroo A, Pereira NL, Bigatello LM, Dec GW, Zapol WM, Bloch KD, Semigran MJ. Hemodynamic effects of sildenafil in patients with congestive heart failure and pulmonary hypertension: combined administration with inhaled nitric oxide. Chest, 2005, 127:1647-53.

- 100. Sebag IA, Handschumacher MD, Ichinose F, Morgan JG, Hataishi R, Rodrigues ACT, Guerrero JL, Steudel W, Raher MJ, Halpern EF, Derumeaux G, Bloch KD, Picard MH, and Scherrer-Crosbie M. Quantitative assessment of regional myocardial function in mice by tissue doppler imaging: comparison with hemodynamics and sonomicrometry. Circulation 2005, 111:2611-6.
- 101. Yu PB, Beppu H, Kawai N, Li E, Bloch KD. BMP type II receptor deletion reveals BMP ligand-specific gain of signaling in pulmonary artery smooth muscle cells. J Biol Chem 2005, 280:24443-24450.
- 102. Lepore JJ, Dec GW, Zapol, WM, Bloch KD, Semigran MJ. Combined administration of intravenous dipyridamole and inhaled nitric oxide to assess reversibility of pulmonary arterial hypertension in potential transplant patients. Journal of Heart and Lung Transplantation 2005, in press.
- 103. Caironi P, Ichinose F, Liu R, Jones RC, Bloch KD, Zapol WM. 5-lipoxygenase deficiency prevents respiratory failure during ventilator-induced lung injury. American Journal of Respiratory and Critical Care Medicine 2005, in press.

Proceedings of Meetings:

1. Bloch KD, Seidman JG, Seidman CE. Structure and expression of the atrial natriuretic factor gene. In: Atrial hormones and other natriuretic factors. Clinical Physiology Series, American Physiological Society, Bethesda, MD, 1987, p. 7-18.

Reviews, Chapters, and Editorials:

- 1. Seidman CE, Bloch KD, Zisfein JB, Smith JA, Haber E, Homey CJ, Duby AD, Choi E, Graham RM, Seidman JG. Molecular studies of the atrial natriuretic factor gene. Hypertension 1985; 7 (part II): 31-34.
- 2. Seidman CE, Bloch KD. Molecular approaches to the study of atrial natriuretic factor. Am J Med Sci 1987; 294:144-149.
- 3. Neer EJ, Kim SY, Ang SL, Bloch DB, Bloch KD, Kawahara K, Tolman C, Lee R, Logethetis D, Kim D, Seidman JG, Clapham DE. Functions of G protein subunits. Cold Spring Harbor Symp. Quant. Biol. 1989; 53:241-246.
- 4. Bloch KD. The nitric oxide synthase gene family. In: Molecular Biology of the Kidney in Health and Disease. Schlondorff D and Bonventre JV, ed., Marcel Dekker Inc. New York, NY, chapter 11, pp 133-142, 1995.
- 5. Lloyd-Jones D, Bloch KD. Vascular biology of nitric oxide and its role in atherogenesis. In: Annual Review of Medicine. C. Coggins, editor. Annual Reviews Inc. Palo Alto CA, volume 47, p. 365-376, 1996.

- 6. Lepore JJ, Bloch KD. Nitric oxide and pulmonary hypertension. In: Loscalzo J and Vita JA, eds. Nitric oxide and the cardiovascular system. Totowa, NJ: Humana Press, Inc., 1999, pp. 247-272.
- 7. Bloch KD. Regulation of endothelial NO synthase mRNA stability: RNA-binding proteins crowd on the 3'-untranslated region. Circ. Res. 1999, 85:653-655.
- 8. Passeri J, Bloch, KD. Nitric Oxide and Cardiac Remodeling. In: Hajjar R, Del Monte F, eds. Heart Failure Clinics. Elsevier Science, Inc. 2005, in press.

Books, Monographs, and Text Books:

1. Zapol WM, Bloch KD, editors. Nitric Oxide and the Lung for the series Lung Biology in Health and Disease. Lenfant C, executive editor. Marcel Dekker Inc. New York, NY, 1996.

Clinical Communications: none

Educational Material: none

Thesis: none

Nonprint Materials: none

Patents:

5,904,938: Treatment of vascular thrombosis and restenosis with inhaled nitric oxide. 6,063,407: Treatment of vascular thrombosis and restenosis with inhaled nitric oxide.

6,183,988: Leukocyte-specific protein and gene, and methods of use thereof. 6,601,580: Enhancing therapeutic effectiveness of nitric oxide inhalation.

6,656,452: Use of inhaled NO as anti-inflammatory agent.

6,720,309: Method of inducing vasodilation and treating pulmonary hypertension using adenoviral-mediated transfer of the nitric oxide synthase gene.

6,811,768: Use of inhaled NO as anti-inflammatory agent.

EXHIBIT B

Ехнівіт В

FIGURE 1

Alignment of human DNMT3A from ATCC Deposit No. 98809 (top) and currently amended SEQ ID NO:3 (bottom)¹

gccgcggcaccagggcgcagccgggccggcccacccggccatacggtggagc	
atcgaagccccacccacaggctgacagaggcaccgttcaccagagggctcaacaccgg	
atctatgtttaagttttaactctcgcctccaaagaccacgataattccttcc	
cagcagcccccagccccgcgcagccccagcctgcctcccggcgcccagatgcccgccaf	
gccctccagcggccccggggacaccagcagctctgctgcggagcgggaggaggaccgaad	
ggacggagagcaggaggagccgcgtggcaaggaggagcgccaagagcccagcaccaggacggagaggag	
ggcacggaaggtggggcggcctgggaggaagcgcaagcaccccccggtggaaagcggtgaggacggaagggaagcgcaagcaccccccggtggaaagcggtgaggcacggaagggaagggaagcgcacgccccggtggaaagcggtgaggaag	
cacgccaaaggaccctgcggtgatctccaagtccccatccat	
ctcagagctattacccaatggggacttggagaagcggagtgagccccagccag	
agggtgcagctgagac	
cctgcctgaagcctcaagagcagtggaaaatggctgctgcacccccaaggagggccgagg 	
agccctgcagaagcgggcaaagaacagaaggagaccaacatcgaatccatgaaaatgga 	

¹ Bolded nucleotides indicate nucleotides that were amended on July 23, 2001.

gggctcccgggggccggctgcggggtggcttgggctgggagtccagcctccgtcagcggcc	780
catgccgaggctcaccttccaggcgggggacccctactacatcagcaagcgcaagcggga :	840
cgagtggctggcacgctggaaaagggaggctgagaagaaagccaaggtcattgcaggaat	900
gaatgctgtggaagaaaaccaggggcccggggagtctcagaaggtggaggaggccagcc	960
tcctgctgtgcagcagcccactgaccccgcatcccccactgtggctaccacgcctgagcc	1020
cgtggggtccgatgctggggacaagaatgccaccaaagcaggcgatgacgagccagagta	1080
cgaggacggccggggctttggcattggggagctggtgtgggggaaactgcggggcttctc	1140
ctggtggccaggccgcattgtgtcttggtggatgacgggccggagccgagcagctgaagg	1200
cacccgctgggtcatgtggttcggagacggcaaattctcagtggtgtgtgt	1260
gatgccgctgagctcgttttgcagtgcgttccaccaggccacgtacaacaagcagcccat	1320
gtaccgcaaagccatctacgaggtcctgcaggtggccagcagccgcgcggggaagctgtt	1380
cccggtgtgccacgacagcgatgagagtgacactgccaaggccgtggaggtgcagaacaa	1440
gcccatgattgaatgggccctggggggcttccagccttctggccctaagggcctggagcc	1500

accagaagaagaagaatccctacaaagaagtgtacacggacatgtgggtgg	
ggcagctgcctacgcaccacctccaccagccaaaaagccccggaagagcacagcggagaa	
gcccaaggtcaaggagattattgatgagcgcacaagagagcggctggtgtacgaggtgcg	
gcagaagtgccggaacattgaggacatctgcatctcctgtgggagcctcaatgttaccct	
ggaacaccccctcttcgttggaggaatgtgccaaaactgcaagaactgctttctggagtg	
tgcgtaccagtacgacgacggctaccagtcctactgcaccatctgctgtgggggccg	
tgaggtgctcatgtgcggaaacaacaactgctgcaggtgcttttgcgtggagtgtgtgga 	
cctcttggtggggccgggggctgcccaggcagccattaaggaagacccctggaactgcta	
catgtgcgggcacaagggtacctacgggctgctgcggcgagaggactggccctcccg	
gctccagatgttcttcgctaataaccacgaccaggaatttgaccctccaaaggtttaccc 	
acctgtcccagctgagaagaggaagcccatccgggtgctgtctctctttgatggaatcgc	
tacagggctcctggtgctgaaggacttgggcattcaggtggaccgctacattgcctcgga	
ggtgtgtgaggactccatcacggtgggcatggtgcggcaccaggggaagatcatgtacgt 	
cggggacgtccgcagcgtcacacagaagcatatccaggagtggggcccattcgatctggt 	
gattgggggcagtccctgcaatgacctctccatcgtcaaccctgctcgcaagggcctcta 	

cgagggcactggccggctcttctttgagttctaccgcctcctgcatgatgcgcggcccaa	2460
ggagggagatgatcgccccttcttctggctctttgagaatgtggtggccatgggcgttag	2520
tgacaagagggacatctcgcgatttctcgagtccaaccctgtgatgattgat	2580
agtgtcagctgcacacagggcccgctacttctggggtaaccttcccggtatgaacaggcc 	2640
gttggcatccactgtgaatgataagctggagctgcaggagtgtctggagcatggcaggat 	2700
agccaagttcagcaaagtgaggaccattactacgaggtcaaactccataaagcagggcaa 	2760
agaccagcattttcctgtcttcatgaatgagaaagaggacatcttatggtgcactgaaat 	2820
ggaaagggtatttggtttcccagtccactatactgacgtctccaacatgagccgcttggc 	2880
gaggcagagactgctgggccggtcatggagcgtgccagtcatccgccacctcttcgctcc 	2940
gctgaaggagtattttgcgtgtgtgtaagggacatgggggcaaactgaggtagcg 	

FIGURE 2

Alignment of predicted amino acids encoded by DNMT3A cDNA in ATCC Deposit No. 98809 (top) and predicted amino acids encoded by currently amended SEQ ID NO:3 (bottom)²

MPAMPSSGPGDTSSSAAEREEDRKDGEEQEEPRGKEERQEPSTTARKVGRPGRKR	55
MPAMPSSGPGDTSSSAAEREEDRKDGEEQEEPRGKEERQEPSTTARKVGRPGRKR	55
KHPPVESGDTPKDPAVISKSPSMAQDSGASELLPNGDLEKRSEPQPEE RVQLRPC	110
KHPPVESGDTPKDPAVISKSPSMAQDSGASELLPNGDLEKRSEPQPEEGSPAGGQ.	110
LKPQEQWKMAAAPPRRAEEPLQKRAKNRRRPTSNP*KWRAPGAGCGVAWAGSPAS	165
KGGAPAEGEGAAETLPEASRAVENGCCTPKEGRGAPAEAGKEQKETNIESMKMEG	165
VSGPCRGSPSRRGTPTTSASASGTSGWHAGKGRLRRKPRSLQE*MLWKKTRGPGS	220
SRGRLRGGLGWESSLRQRPMPRLTFQAGDPYYISKRKRDEWLARWKREAEKKAKV	220
LRRWRRPALLLCSSPLTPHPPLWLPRLSPWGPMLGTRMPPKQAMTSQSTRTAGAL	275
GMNAVEENQGPGESQKVEEASPPAVQQPTDPASPTVATTPEPVGSDAGDKNIAAT	275
ALGSWCGGNCGASPGGQAALCLGG*RAGAEQLKAPAGSCGSETANSQWCVLRS*C	330
KAGDDEPEYEDGRGFGIGELVWGKLRGFSWWPGRIVSWWMTGRSRAAEGTRWVMW	330
R*ARFAVRSTRPRTTSSPCTAKPSTRSCRWPAAARGSCSRCATTAMRVTLPRPWR	385
FGDGKFSVVCVEKLMPLSSFCSAFHQATYNKQPMYRKAIYEVLQVASSRAGKLFP	385
CRTSP*LNGPWGASSLLALRAWSHQKKRRIPTKKCTRTCGWNLRQLPTHHLHQPK	440
VCHDSDESDTAKAVEVQNKPMIEWALGGFQPSGPKGLEPPEEEKNPYKEVYTDMW	440
SPGRAQRRSPRSRRLLMSAQESGWCTRCGRSAGTLRTSASPVGASMLPWNTPSSL	495
VEPEAAAYAPPPPAKKPRKSTAEKPKVKEIIDERTRERLVYEVRQKCRNIEDICI	495
EECAKTARTAFWSVRTSTTTTATSPTAPSAVGAVRCSCAETTTAAGAFAWSVWTS	550
SCGSLNVTLEHPLFVGGMCQNCKNCFLECAYQYDDDGYQSYCTICCGGREVLMCG	550
WWGRGLPRQPLRKTPGTATCAGTRVPTGCCGGERTGPPGSRCSSLITTTRNLTLQ	605
NNNCCRCFCVECVDLLVGPGAAQAAIKEDPWNCYMCGHKGTYGLLRRREDWPSRL	605
RFTHLSQLRRGSPSGCCLSLMESLQGSWC*RTWAFRWTATLPRRCVRTPSRWAWC	660
QMFFANNHDQEFDPPKVYPPVPAEKRKPIRVLSLFDGIATGLLVLKDLGIQVDRY	660
GTRGRSCTSGTSAASHRSISRSGAHSIW*LGAVPAMTSPSSTLLARASTRALAGS	715
IASEVCEDSITVGMVRHQGKIMYVGDVRSVTQKHIQEWGPFDLVIGGSPCNDLSI	715
SLSSTASCMMRGPRREMIAPSSGSLRMWWPWALVTRGTSRDFSSPTL**LMPKKC	770
VNPARKGLYEGTGRLFFEFYRLLHDARPKEGDDRPFFWLFENVVAMGVSDKRDIS	770
QLHTGPATSGVTFPV*TGRWHPL*MISWSCRSVWSMAG*PSSAK*GPLLRGQTP*	825
RFLESNPVMIDAKEVSAAHRARYFWGNLPGMNRPLASTVNDKLELQECLEHGRIA	825

 $^{^2}$ * indicates a predicted stop codon. Bolded amino acids are encoded by nucleotides located downstream of the deletion.

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SRAKTSIFLSS*MRKRTSYGALKWKGYLVSQSTILTSPT*AAWRGRDCWAGHGAC	880
KFSKVRTITTRSNSIKQGKDQHFPVFMNEKEDILWCTEMERVFGFPVHYTDVSNM	880
QSSATSSLR*RSILRVCKGHGGKLR**AAWRG	912
SRLARORLLGRSWSVPVIRHLFAPLKEYFACV	912

FIGURE 3

Human DNMT3A from ATCC Deposit No. 98809 with forward reading frames (DNMT3A amino acid residues bolded)

+3: +2:	R G T R A R S R A G P T P P A I R P R H Q G A Q P G R P D P T G H T A A A P G R A A G P A R P H R P Y	51
+3: +2:	GGTGGAGCCATCGAAGCCCCACCCACAGGCTGACAGAGGCACCGTTCACC WSHRSPHPQADRGTVHQ VEPSKPPPTG*QRHRSP GGAIEAPTHRLTEAPFT	102
+3: +2:	AGAGGGCTCAACACCGGGATCTATGTTTAAGTTTTAACTCTCGCCTCCAAA R A Q H R D L C L S F N S R L Q R E G S T P G S M F K F * L S P P K R G L N T G I Y V * V L T L A S K	153
+3: +2:	GACCACGATAATTCCTTCCCCAAAGCCCAGCAGCCCCCAGCCCCGCGCAG PR*FLPQSPAAPQPRAAA TTIIPSPKPSSPPAPRS DHDNSFPKAQQPPSPAQ	204
+3: +2:	CCCCAGCCTGCCTCCGGCGCCCAGATGCCCTCCAGCGGCCC P A C L P A P R C P P C P P A A P P S L P P G A Q M P A M P S S G P P Q P A S R R P D A R H A L Q R P	255
DNA: +3: +2: +1:	CGGGGACACCAGCAGCTCTGCTGCGGAGCGGAGGAGGACCGAAAGGACGG G T P A A L L R S G R R T E R T E G D T S S S A A E R E E D R K D G R G H Q Q L C C G A G G G P K G R	306
DNA: +3: +2: +1:	AGAGGAGCAGGAGCCCCGCGTGGCAAGGAGCGCCAAGAGCCCAGCAC R S R R S R V A R R S A K S P A P E E Q E E P R G K E E R Q E P S T R G A G G A A W Q G G A P R A Q H	357
DNA: +3: +2: +1:	CACGGCACGGAAGGTGGGGCGGCCTGGGAGGAAGCGCAAGCACCCCCGGT R H G R W G G L G G S A S T P R W T A R K V G R P G R K R K H P P V H G T E G G A A W E E A Q A P P G	408
+3:	GGAAAGCGGTGACACGCCAAAGGACCCTGCGGTGATCTCCAAGTCCCCATC K A V T R Q R T L R * S P S P H P E S G D T P K D P A V I S K S P S G K R * H A K G P C G D L Q V P I	459
DNA: +3: +2: +1:	CATGGCCCAGGACTCAGGCGCCTCAGAGCTATTACCCAATGGGGACTTGGA W P R T Q A P Q S Y Y P M G T W R M A Q D S G A S E L L P N G D L E H G P G L R R L R A I T Q W G L G	510
DNA: +3: +2: +1:	GAAGCGGAGTGAGCCCCAGCCAGAGGAG. S G V S P S Q R R. K R S E P Q P E E. DELETION. E A E * A P A R G E.	561

	610
DNA: GCS*DPA*S +3: RVQLRPCLKP +2: GAAETLPEA	612
DNA: CTCAAGAGCAGTGGAAAATGGCTGCTGCACCCCCAAGGAGGGCCGAGGAGC +3: L K S S G K W L L H P Q G G P R S +2: Q E Q W K M A A A P P R R A E E P +1: S R A V E N G C C T P K E G R G A	663
DNA: CCCTGCAGAAGCAGAAGAACAGAAGGAGACCAACATCGAATCCATGAA +3: P C R S G Q R T E G D Q H R I H E +2: L Q K R A K N R R R P T S N P * K +1: P A E A G K E Q K E T N I E S M K	714
DNA: AATGGAGGGCTCCCGGGGGCCGGCTGCGGGGTGGCTTGGGCTGGGAGTCCAG +3: N G G L P G P A A G W L G L G V Q +2: W R A P G A G C G V A W A G S P A +1: M E G S R G R L R G G L G W E S S	765
DNA: CCTCCGTCAGCGGCCCATGCCGAGGCTCACCTTCCAGGCGGGGGACCCCTA +3: P P S A A H A E A H L P G G G P L +2: S V S G P C R G S P S R R G T P T +1: L R Q R P M P R L T F Q A G D P Y	816
DNA: CTACATCAGCAAGCGCAAGCGGGACGAGTGGCTGGCACGCTGGAAAAGGGA +3: L H Q Q A Q A G R V A G T L E K G +2: T S A S A S G T S G W H A G K G R +1: Y I S K R K R D E W L A R W K R E	867
DNA: GGCTGAGAAGAAGCCAAGGTCATTGCAGGAATGAATGCTGTGGAAGAAAA +3: G * E E S Q G H C R N E C C G R K +3: G * E E S Q G H C R N E C C G R K +2: L R R K P R S L Q E * M L W K K T +1: A E K K A K V I A G M N A V E E N	918
DNA: CCAGGGGCCCGGGGAGTCTCAGAAGGTGGAGGAGGCCAGCCCTCCTGCTGT +3: P G A R G V S E G G G Q P S C C +2: R G P G S L R R W R R P A L L L C +1: Q G P G E S Q R V E E A S P P A V	969
DNA: GCAGCAGCCCACTGACCCCGCATCCCCCACTGTGGCTACCACGCCTGAGCC +3: A A A H * P R I P H C G Y H A * A +2: S S P L T P H P P L W L P R L S P	1020
DNA: CGTGGGGTCCGATGCTGGGGACAAGAATGCCACCAAAGCAGGCGATGACGA +3: R G V R C W G Q E C H Q S R R * R +2: W G P M L G T R M P P K Q A M T S +1: V G S D A G D K N A T K A G D D E	1071
DNA: GCCAGAGTACGAGGACGGCGGGCTTTGGCATTGGGGAGCTGGTGTGGGG +3: A R V R G R P G L W H W G A G V G +2: Q S T R T A G A L A L G S W C G G +1: P E Y E D G R G F G I G E L V W G	1122
DNA: GAAACTGCGGGGCTTCTCCTGGTGGCCAGGCCGCATTGTGTTGTTGGTGGAT +3: E T A G L L V A R P H C V L V D +2: N C G A S P G G Q A A L C L G G * +1: K L R G F S W W P G R I V S W W M	1173

+3: +2:,	GACGGGCCGAGCAGCAGCTGAAGGCACCCGCTGGGTCATGTGGTTCGG D G P E P S S * R H P L G H V V R R A G A E Q L K A P A G S C G S E T G R S R A A E G T R W V M W F G	1224
+3: +2:	AGACGGCAAATTCTCAGTGGTGTGTGTGAGAAGCTGATGCCGCTGAGCTC R R Q I L S G V C * E A D A A E L T A N S Q W C V L R S * C R * A R D G K F S V V C V E K L M P L S S	1275
+3: +2:	GTTTTGCAGTGCGTTCCACCAGGCCACGTACAACAAGCAGCCCATGTACCG V L Q C V P P G H V Q Q A A H V P F A V R S T R P R T T S S P C T A F C S A F H Q A T Y N K Q P M Y R	1326
+3:	CAAAGCCATCTACGAGGTCCTGCAGGTGGCCAGCAGCCGCGGGGAAGCT Q S H L R G P A G G Q Q P R G E A K P S T R S C R W P A A A R G S C K A I Y E V L Q V A S S R A G K L	1377
+3: +2:	GTTCCCGGTGTGCCACGACAGCGATGAGAGTGACACTGCCAAGGCCGTGGA V P G V P R Q R * E * H C Q G R G S R C A T .T A M R V T L P R P W R F P V C H D S D E S D T A K A V E	1428
+3: +2:	GGTGCAGAACAAGCCCATGATTGAATGGGCCCTTGGGGGGGCTTCCAGCCTTC G A E Q A H D * M G P G G L P A F C R T S P * L N G P W G A S S L L V Q N K P M I E W A L G G F Q P S	1479
+3: +2:	TGGCCCTAAGGGCCTGGAGCCACCAGAAGAAGAAGAATCCCTACAAAGA W P * G P G A T R R R E E S L Q R A L R A W S H Q K K R R I P T K K G P K G L E P P E E E K N P Y K E	1530
+3:	AGTGTACACGGACATGTGGGTGGAACCTGAGGCAGCTGCCTACGCACCACC S V H G H V G G T * G S C L R T T C T R T C G W N L R Q L P T H H L V Y T D M W V E P E A A A Y A P P	1581
+3:	TCCACCAGCCAAAAAGCCCCGGAAGAGCACAGCGAGAAGCCCAAGGTCAA S T S Q K A P E E H S G E A Q G Q H Q P K S P G R A Q R R S P R S R P P A K K P R K S T A E K P K V K	1632
+3: +2:	GGAGATTATTGATGAGCGCACAAGAGAGCGGCTGGTGTACGAGGTGCGGCA G D Y * * A H K R A A G V R G A A R L L M S A Q E S G W C T R C G R E I I D E R T R E R L V Y E V R Q	1683
+3:	GAAGTGCCGGAACATTGAGGACCATCTCCATCTCCTGTGGGAGCCTCAATGT E V P E H * G H L H L L W E P Q C S A G T L R T S A S P V G A S M L K C R N I E D I C I S C G S L N V	1734
+3:	TACCCTGGAACACCCCCTCTTCGTTGGAGGAATGTGCCAAAACTGCAAGAA Y P G T P P L R W R N V P K L Q E P W N T P S S L E E C A K T A R T T L E H P L F V G G M C Q N C K N	1785

+3: +2:	CTGCTTTCTGGAGTGTGCGTACCAGTACGACGACGACGGCTACCAGTCCTA L L S G V C V P V R R R L P V L A F W S V R T S T T T T A T S P T C F L E C A Y Q Y D D D G Y Q S Y	1836
+3: +2:	CTGCACCATCTGCTGTGGGGGCCGTGAGGTGCTCATGTGCGGAAACAACAA L H H L L W G P * G A H V R K Q Q A P S A V G A V R C S C A E T T T C T I C C G G R E V L M C G N N N	1887
+3: +2:	CTGCTGCAGGTGCTTTTGCGTGGAGTGTGTGGACCTCTTGGTGGGGCCGGG L L Q V L L R G V C G P L G G A G A A G A F A W S V W T S W W G R G C C R C F C V E C V D L L V G P G	1938
+3: +2:	GGCTGCCCAGGCAGCCATTAAGGAAGACCCCTGGAACTGCTACATGTGCGG G C P G S H * G R P L E L L H V R L P R Q P L R K T P G T A T C A G A A Q A A I K E D P W N C Y M C G	1989
+3: +2:	GCACAAGGGTACCTACGGGCTGCTGCGGCGGCGAGAGGACTGGCCCTCCCG A Q G Y L R A A A A A R G L A L P T R V P T G C C G G E R T G P P G H K G T Y G L L R R R E D W P S R	2040
+3: +2:	GCTCCAGATGTTCTTCGCTAATAACCACGACCAGGAATTTGACCCTCCAAA A P D V L R * * P R P G I * P S K S R C S S L I T T T R N L T L Q R L Q M F F A N N H D Q E F D P P K	2091
+3: +2:	GGTTTACCCACCTGTCCCAGCTGAGAAGAGGAAGCCCATCCGGGTGCTGTC G L P T C P S * E E E A H P G A V F T H L S Q L R R G S P S G C C L V Y P P V P A E K R K P I R V L S	2142
+3: +2:	TCTCTTTGATGGAATCGCTACAGGGCTCCTGGTGCTGAAGGACTTGGGCAT S L * W N R Y R A P G A E G L G H S L M E S L Q G S W C * R T W A F L F D G I A T G L L V L K D L G I	2193
+3: +2:	TCAGGTGGACCGCTACATTGCCTCGGAGGTGTGTGAGGACTCCATCACGGT S G G P L H C L G G V * G L H H G R W T A T L P R R C V R T P S R W Q V D R Y I A S E V C E D S I T V	2244
+3: +2:	GGGCATGGTGCGGCACCAGGGGAAGATCATGTACGTCGGGGACGTCCGCAG G H G A A P G E D H V R R G R P Q A W C G T R G R S C T S G T S A A G M V R H Q G K I M Y V G D V R S	2295
+3: +2:	CGTCACAGAAGCATATCCAGGAGTGGGGCCCATTCGATCTGGTGATTGG R H T E A Y P G V G P I R S G D W S H R S I S R S G A H S I W * L G V T Q K H I Q E W G P F D L V I G	2346
DNA: +3: +2: +1:	GGGCAGTCCCTGCAATGACCTCTCCATCGTCAACCCTGCTCGCAAGGGCCT G Q S L Q * P L H R Q P C S Q G P A V P A M T S P S S T L L A R A S G S P C N D L S I V N P A R K G L	2397

+3: +2:	CTACGAGGGCACTGGCCGGCTCTTCTTTGAGTTCTACCGCCTCCTGCATGA L R G H W P A L L * V L P P P A * T R A L A G S S L S S T A S C M M Y E G T G R L F F E F Y R L L H D	2448
+3: +2:	TGCGCGGCCCAAGGAGGGAGATGATCGCCCCTTCTTCTGGCTCTTTGAGAA C A A Q G G R * S P L L L A L * E R G P R R E M I A P S S G S L R M A R P K E G D D R P F F W L F E N	2499
+3: +2:	TGTGGTGGCCATGGGCGTTAGTGACAAGAGGGACATCTCGCGATTTCTCGA C G G H G R * * Q E G H L A I S R W W P W A L V T R G T S R D F S S V V A M G V S D K R D I S R F L E	2550
+3: +2:	GTCCAACCCTGTGATGATTGATGCCAAAGAAGTGTCAGCTGCACACAGGGC V Q P C D D * C Q R S V S C T Q G P T L * * L M P K K C Q L H T G P S N P V M I D A K E V S A A H R A	2601
+3: +2:	CCGCTACTTCTGGGGTAACCTTCCCGGTATGAACAGGCCGTTGGCATCCAC P L L L G * P S R Y E Q A V G I H A T S G V T F P V * T G R W H P L R Y F W G N L P G M N R P L A S T	2652
+3: +2:	TGTGAATGATAAGCTGGAGCTGCAGGAGTGTCTGGAGCATGGCAGGATAGC C	2703
+3: +2:	CAAGTTCAGCAAAGTGAGGACCATTACTACGAGGTCAAACTCCATAAAGCA Q V Q Q S E D H Y Y E V K L H K A S S A K * G P L L R G Q T P * S R K F S K V R T I T T R S N S I K Q	2754
+3:+2:	GGGCAAAGACCAGCATTTTCCTGTCTTCATGAATGAGAAAGAGGACATCTT GQRPAFSCLHEKERGHL AKTSIFLSS*MRKRTSY GKDQHFPVVFMNEKEDIL	2805
+3: +2:	ATGGTGCACTGAAATGGAAAGGGTATTTGGTTTCCCAGTCCACTATACTGA M V H * N G K G I W F P S P L Y * G A L K W K G Y L V S Q S T I L T W C T E M E R V F G F P V H Y T D	2856
+3: +2:	CGTCTCCAACATGAGCCGCTTGGCGAGGCAGAGACTGCTGGGCCGGTCATG R L Q H E P L G E A E T A G P V M S P T * A A W R G R D C W A G H G V S N M S R L A R Q R L L G R S W	2907
+3: +2:	GAGCGTGCCAGTCATCCGCCACCTCTTCGCTCCGCTGAAGGAGTATTTTGC E R A S H P P P L R S A E G V F C A C Q S S A T S S L R * R S I L R S V P V I R H L F A P L K E Y F A	2958
+3: +2:	GTGTGTGTAAGGGACATGGGGGCAAACTGAGGTAGCG V C V R D M G A N * G S V C K G H G G K L R * C V * G T W G Q T E V A	2995

EXHIBIT C

LETTER

Estimation of Errors in "Raw" DNA Sequences: A Validation Study

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As DNA sequencing is performed more and more in a mass-production-like manner, efficient quality control measures become increasingly important for process control, but so also does the ability to compare different methods and projects. One of the fundamental quality measures in sequencing projects is the position-specific error probability at all bases in each individual sequence. Accurate prediction of base-specific error rates from 'raw' sequence data would allow immediate quality control as well as benchmarking different methods and projects while avoiding the inefficiencies and time delays associated with resequencing and assessments after "finishing" a sequence. The program PHRED provides base-specific quality scores that are logarythmically related to error probabilities. This study assessed the accuracy of PHRED's error-rate prediction by analyzing sequencing projects from six different large-scale sequencing laboratories. All projects used four-color fluorescent sequencing, but the sequencing methods used varied widely between the different projects. The results indicate that the error-rate predictions such as those given by PHRED can be highly accurate for a large variety of different sequencing methods as well as over a wide range of sequence quality.

In DNA sequencing, knowledge about the accuracy of sequences can be very valuable. For example, different large-scale sequencing projects may produce sequences at similar rates and costs but with significantly different error rates in the final sequence. One major determinant in the final error rate is the accuracy of the "raw" sequence. Knowledge about the frequency and location of errors in the raw sequence data can help to direct "polishing" efforts to the places where additional effort is needed; it also enables the comparison between different sequencing projects without requiring that the same region be sequenced in each project.

Another area where estimates about sequence error rates would be beneficial is technology development. Accurate error estimates at each base would enable "quality benchmarking" hetween different methods, thus enabling researchers to choose the method that fills their needs for accuracy and throughput best.

Several groups have developed mathematical models to predict the error probability at any given position in raw sequences. Lawrence and Solovyev used linear discriminant analysis to calculate separate probability estimates for insertions, deletions, and mismatches (Lawrence and Solovyev 1994). Ewing and Green (1998) developed the program

PHRED, which calculates a quality score at each base. This quality score q is logarithmically linked to the error probability p: $q = -10 \times \log_{10} (p)$ (for a discussion of how quality scores are calculated and what the limitations are, see Ewing et al. (1998). When used in combination with sequence assembly and finishing programs that utilize these error estimates, reliable error probabilities promise to increase the accuracy of consensus sequences and to reduce the efforts required in the finishing phase of sequencing projects (Churchill and Waterman 1992; Bonfield and Staden 1995).

To examine the accuracy of probability estimates made by the program PHRED, we compared the actual and predicted error rates for six different cosmid- or BAC-sized projects that were produced by six different large-scale sequencing centers in the United States. All of these six projects used four-color fluorescent sequencing machines; however, the DNA preparation methods, sequencing enzymes, fluorescent dyes and chemistries, and gel lengths varied significantly between the six groups. Table 1 gives an overview of the sequencing projects analyzed. Table 2 lists the different methods used.

RESULTS

Error Rate Prediction Accuracy for Six Projects

A comparison of actual and predicted error rates for the six projects in this study is shown in Table 3.

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		Averag		
Project	Reads	Aligned bases	aligned read length	
A B	455 1277	416,214 871,230	915 682	
C D E	1065 834 1638	603,655 414,595	567 497	
F Fotal	1885 7154	1,149,209 907,796 4,362,699	702 482 610	

The results indicate that PHRED is very successful in identifying bases with low error probabilities. For example, the 1.28 million bases with quality scores of 4-12 (corresponding to error probabilities between 39.8% and 6.3%) contain a lotal of 187,926 errors. In contrast, the 1.44 million bases with quality scores between 33 and 42 (corresponding to error probabilities between 0.05% and 0.006%) contain only 237 errors, which translates into a 790-fold lower error rate. The trend toward lower error rates can also be observed for each individual project. In most cases, the actual number of errors is close to the predicted error rate: It is also apparent that the actual error rate is typically lower than the predicted error rate.

Both the high overall accuracy and the tendency to slightly overpredict errors are confirmed by statistical analysis, as shown in Table 4. The correlation between predicted and actual error frequen-Cles is excellent for all projects (Spearman correlafrom coefficient >0.89, P < 0.0001). Averaged over all projects, the actual error rate is 84.5% of the predicted error rate; the slope of the relation between predicted and actual error rates differs slightly between projects and ranges from 76.6% to 88,4%. To put these differences between projects in relation, it is worthwhile remembering that PIIRED quality scores cover a wide dynamic range: The maximum quality score of 51 corresponds to a 50,000 fold lower predicted error rate than the minimum qual-Ity score of 4. Even the relative difference between successive quality is larger than the relative difference in the slopes; for example, a quality score of 10 corresponds to an error probability of 10%, whereas a score of 9 corresponds to an error probability of

A different way of looking at the relation between the actual and predicted error rates is shown

in Figure 1. Here, the error rates as a function of the position within all reads in each of the projects, averaged over 50-base windows, is depicted. For all six projects, the predicted error rates are very close to the actual error rates over the entire length of the sequences. Each project has a characteristic distribution of error rates, which differs from each of the other projects. The minimum error rate differs dramatically between projects. The best projects achieve raw error rates of 0.23%—0.36% in the best region of the sequence read, typically from base 150 to 200. The worst project in the data set had an ~10-fold higher error rate of 2.58%.

Toward the end of sequence reads, the error rates increase and start to exceed 10% between bases 300 and 700. In projects that used mainly short gels (e.g., projects D and F), this increase begins sooner, whereas projects that use longer gels show a markedly longer stretch of low error rates (e.g., projects A and B).

Table 5 summarizes key results for the six projects. The first four projects have similar minimum and average error rates. However, the length of the region where the error rate is below 5% differs significantly, from 403 to 682 bases. The project with the shorter low error rate regions contained larger portions of reads generated on short gels, whereas projects A and B were run exclusively on long gels (ABI373 stretch or ABI377 sequencers). Other factors contributing to differences between the first four projects were differences in sequencing chemistries, production scale, and electrophoresis conditions and machines.

Project E and, in particular, project F, had significantly higher error rates than the first four projects. In projects E and F, every sequence generated for the project had been included in the data set, whereas the other four projects had eliminated some "bad" sequences through manual or auto-

Table 2. Overview of Sequencing Methods Used in the Different Projects

Aben III THE DE	fferent Projects
Template DNA	single-stranded M13,
Sequencing	double-stranded plasmids
enzymes	Sequenase, <i>Taq</i> , KlenTaqTR, AmpliTag FS
Sequencing	Dude primar Area dire
chemistries	Dyes primer (two different dyes chemistrics), dye terminalor
Sequencing	ABI 373, ABI 373 stretch,
machines	ABI 377
Gel length	Only short gels, only long gels, mixes of short and long gels

Table 3. Comparison of Predicted and Actual Error Rates for Six Different Sequencing Projects						
Project	Quality score	4-12	13-22	23-32	33-42	43-51
A	aligned bases	119,246	75,293	70,391	144,876	73,234
	expected errors	20,256	2,064	172	37	1
	actual errors	16,784	1,758	127	17	1
В	aligned bases	182,034	137,940	181,998	399,690	140,176
	expected errors	29,953	3,704	410	102	3
	actual errors	26,038	2,536	287	35	0
C	aligned bases	139,345	131,419	151,197	292,070	68,529
	expected errors	22,277	3,411	357	74	2
	actual errors	16,670	1,513	194	26	3
D	aligned bases	103,898	68,995	68,613	153,730	111,752
	expected errors	16,880	1,919	168	38	3
	actual errors	14,495	1,924	146	59	2
E .	aligned bases	378,755	217,438	167,968	392,717	144,313
	expected errors	63,947	6,336	418	95	4
	actual errors	55,968	6,516	355	67	5
: 	aligned bases expected errors actual errors	359,809 66,938 57,971	136,688 4,079 3,856	98,840 256 332	64,035 23 33	5,130 0 1
All	aligned bases	1,283,087	767,773	739,007	1,447,118	543,134
	expected errors	220,252	21,513	1,781	370	13
	actual errors	187,926	18,103	1,441	237	12

matic inspection. After eliminating <10% of the worst sequences in project E, the error rates for the remaining sequences were comparable to those of the first four projects. In contrast, project E showed a much more uniform distribution of sequence quality.

The last column in Table 5 shows the average number of bases with an estimated error probability of at most 0.1%, which is equivalent to a quality score of at least 30. The count of such "very high-quality" bases is a good indicator of sequence quality, both for individual sequences and, when aver-

Table 4. Summary of Statistical Analysis Results					garan a sanggaga
Project	Spearman P	P > p	Slope	t ratio	P > t
A	0.9646	< 0.0001	0.818	75.1	<0.0001
В	0.9890	< 0.0001	0.874	98.2	< 0.0001
C	0.9846	< 0.0001	0.766	71.6	< 0.0001
.D °	0.8692	< 0.0001	0.855	68.3	< 0.0001
Έ	0.9956	< 0.0001	0.884	144.3	< 0.0001
Ė,	0.9968	< 0.0001	0.865	151.6	< 0.0001
All	0.9964	<0.0001	0.845	174.5	< 0.0001

In project D, the Spearman correlation coefficient p was artificially low as only very few bases (10) bases had a quality score of 5, and none of these bases contained an actual error (expected: 3.16 errors). Exclusion of this quality score gave a Spearman correlation coefficient of 0.9786 (P < 0.0001). The frequencies in the slope calculations were weighted by the number of bases at any given quality score and, thus, were not sensitive to such small sample distortions (see Methods).

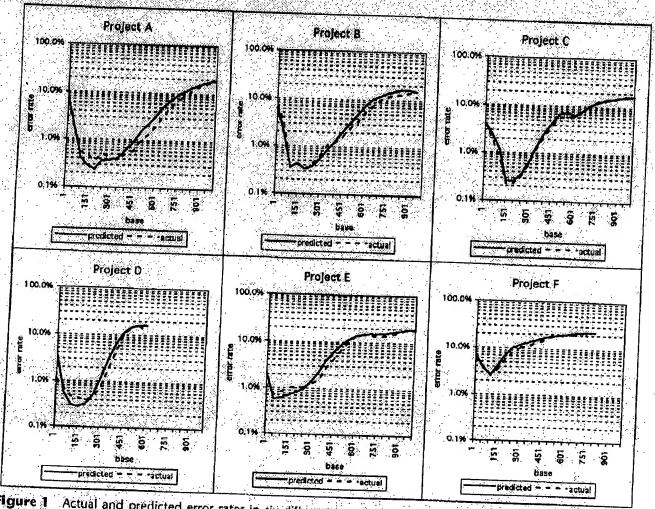


Figure 1 Actual and predicted error rates in six different sequencing projects. Actual error rates and predicted error rates in 50-base windows over the length of the sequence reads, averaged over all reads that could be aligned to the consensus sequence by CROSS_MATCH, are shown. The numbers on the x-axis show the first base in a given 50-base window.

aged over all sequences in a project, as an indicator for the critic project. Compared to the estimated error rates, the count of very high-quality bases is less prone to distortions from a small number of low-quality reads, as the data for project E demonstrate.

Prediction Accuracy for Data Subsets of Different Quality

The quality of sequences within any given project can vary substantially, and the use of predicted error rates has the potential to be a powerful tool for qual-

Project	Actual minimum error rate (%)	Actual average error rate (%)	Six Different Sequent Length of <1% error region	Length of <5% error region	Average bases with
A B C D E I	0.36 0.34 0.23 0.39 0.71 2.58	3.6 2.8 2.4 3.1 4.7 9.2	422 274 291 300 129 0	682 567 479 403 464 162	P(error) <0.1% 468 395 348 294 317 79

ity analysis and control in large-scale DNA sequencing projects. To analyze how accurate PHRED error estimates are for different quality sequences within the same sequencing project, we subdivided a data set into four quartiles, based on the number of very high-quality bases in each sequence (see Methods). The comparison of actual and predicted error rates is shown in Figure 2.

When measured by the error rate in the best region of a sequence, the data quality in the different quartiles varies >100 fold between the best and the worst 25% of the sequences. The best quartile showed ~0.03% error for >100 bases, whereas the error rate in the worst quartile always exceeded 5%. In quartiles 2 and 3, the predicted error rates match the actual error rates very closely. In the best and

worst quartiles, PHRED's accuracy was somewhat lower from base 100 to 500. In the best sequences, PHRED's error estimates were about twofold too high; in the worst sequences, the error estimates were too low, again by a factor of 2. This underprediction of errors can be partially explained by the fact that PHRED gives ambiguous base calls (N's) a quality score of 4, corresponding to an error probability of 39.8%; however, N's will always show up as an actual error. Even in the worst and best quartiles, however, the predicted error rate curves are very similar to the actual error rate curves.

The results shown in Figure 2 also demonstrate that the count of very high-quality bases, or bases with an estimated error probability of at most 0.1%, can be used effectively to characterize the overall

quality of a sequence read. Sorting the sequence reads into quartiles based on the number of very high-quality bases worked well, as shown by the >100-fold difference in the minimum error rate between the first and the fourth quartile.

Other methods to characterize the overall quality of individual reads based on PHRED quality scores can give similar results. For example, counting bases above a minimum quality threshold anywhere in the range of 20-40 gave similar results for most data sets (not shown), and such counts are used by a number of different laboratories as quality measures. Alternatively, the quality values can be converted to error probabilities and averaged to give the predicted error rate for the trace, or summed to give the total predicted number of errors in a trace. However, such averages and totals can sometimes give a misleading picture, as the following example illustrates. Assume that two sequence reads have very similar quality in the alignable part of the read but that one of the two sequences was run much longer and

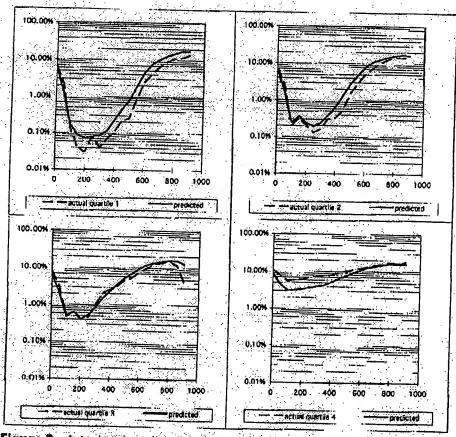


Figure 2 Actual and predicted error rates in different quality subsets of project B. Sequence reads were sorted by the number of bases with a predicted error rate of at most 0.1% (very high-quality bases), and assigned to quartiles, with quartile 1 corresponding to the highest numbers. Actual and predicted error rates for all sequences in each subset were calculated as in Fig. 1. Note that a number of sequence reads that had been rejected because of too low quality were added back to the data set for illustrative purposes, all of which are in quartile 4. These sequences were not included in the data sets used to generate Figs. 1 and 3 and Tables 1 and 3.

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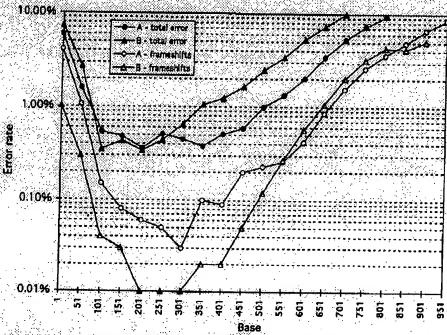


Figure 3 Actual frameshift and total error rates for projects A and B. To calculate frameshift error rates, only insertions and deletions were counted. Mismatch errors, which account for the vast majority of errors after base 150, were included only in the total error count. Note that project B (\spadesuit , \triangle) has a slightly similar or slightly higher total error rate compared to project A (\spadesuit , \bigcirc) but only about one-third as many insertions and deletions up to base 500. For both projects, the frameshift error rate in the raw data is <1 in 1000 for >300 bases, and <1 in 10,000 for >100 bases in project B.

therefore contains a longer unalignable "tail" of very low-quality bases. When calculating the average error rate for these two sequences, the second sequence will have a much higher average error and, therefore, appear to be of lower quality. In contrast, the counts of very high-quality bases for both sequences will be very similar; as the unalignable tails contain few, if any, high-quality bases. Therefore, counts of bases above a high chough quality threshold will give a more robust and clearer picture of trace quality.

Frameshift Error Rates for Different Sequencing Chemistries

Depending on how biologists use DNA sequences, knowledge about total error rates in raw sequences may or may not be sufficient. For example, frame-shift errors in coding sequences will generally lead to incorrectly predicted open reading frame, whereas mismatch errors will do so only if the mismatch introduces a stop codon or a new splice site. At the time of this writing, PHRED did not differentiate between mismatch and frameshift errors, but only estimated total error rates. This might occa-

sionally lead to questionable conclusions, as the results shown in Pigure 3illustrate.

Figure 3 shows the total actual error rates and the frameshift error rates for two projects, A and B. The total erfor rates for both projects are similar for up to 350 bases; after 350 bases, project B has a somewhat higher total error rate. However, examining the frameshift error rate gives rise to a different picture: from base 1 to 500, project A has approximately four times as many insertions and deletions as project B. This difference in frameshift error rates. can be explained by the sequencing chemistries that were used in the two projects. Project B, with the lower frameshift error rate, used only dye terminator chemistry, which is known to eliminate band spacing artifacts from hairpin structures ("compressions"). Project A, on the

other hand, used dye primer chemistry, which is more prone to insertion and deletion errors from mobility artifacts, for most sequencing reactions.

DISCUSSION

As large-scale DNA sequencing has become a more routine and common process, the traditional methods for assessing sequence quality have become unsatisfactory. In projects like single-pass cl)NA sequencing, it is not possible to calculate and compare error rates after linishing a sequence, as finishing never takes place. Even when a comparison between raw and finished sequence can be done, the time delay between raw data generation and quality assessment is often large. This delay makes it difficult to improve ongoing projects, and it sometimes makes it impossible to capture problems early on. Some immediate quality feedback can be reached by including known standard sequences for quality control. However, this approach can be costly, and it fails when error profiles differ between standard and unknown sequences.

In contrast to these traditional methods to assess sequence accuracy, direct estimation of error rates in raw sequence data would enable immediate quality control and feedback. Accurate, base-by-base estimates of error probabilities could also increase the utility of single-pass sequences significantly, allow efficient comparison and optimization of different sequence chemistries, and enable the development of better software tools for sequence assembly and analysis.

The critical question for any error rate prediction tool is how accurate are the error rate estimates, In particular if different sequencing methods and chemistries are used? The results presented herein provide an answer to this question for the program PHRED, as well as clues where further development would be useful. As shown in Tables 3 and 4 and in Figure 1, the agreement between predicted and actual error rates was very good in each of the six different projects analyzed. The observed high level of prediction accuracy in all of these projects is almost asionishing if one takes into account that actual errors are binary (a base is either correct or wrong), whereas predicted error rates are probabilities on a scale from 0.0 to 1.0. The observed tendency to overpredict error rates can be at least partially explained by the "small sample correction" that was used in the derivation of threshold paramcters for quality scores (Ewing and Green 1998). For most practical applications, such a sumewhat conservative estimation of quality scores is tolerable or even desirable. Overall, the results clearly show that error probabilities given by PHRED accurately describe raw sequence data quality.

In judging the usefulness of predicted error probabilities, it is important to know how differences in sequencing methods will influence the prediction accuracy. For example, the larger variation in peak heights tends to be larger in dye terminator sequencing than in dye primer sequencing, and different sequencing enzymes are known to produce different specific height variation patterns. Any estimation of error probabilities that takes the peculiarities of a specific sequencing chemistry into account would therefore be expected to be less accurate for different chemistries.

The projects included in this study were specifically chosen to provide an initial answer to the question of how generally useful PHRED quality scores are. These projects represent the vast majority of different multicolor fluorescent sequencing methods used in the last 3 years: different template DNAs and DNA preparation methods, different enzymes, gel lengths, run conditions, and different fluorescent dyes. The data also include a considerable spread in data quality, both between projects

and within individual projects. None of the projects analyzed here were included in PHRED's training set, and just one of the six laboratories that contributed data to this study also contributed data to the training data sets. One of the projects in this sludy consisted entirely of dye terminator sequences, which presented only a small fraction of the sequences in the test data set. Another project exclusively used a set of fluorescent dyes different from those used in the training sets. Fach project differed from the other projects in this study in at least one, and typically many, experimental aspects like template preparation, sequencing enzymes, gel run conditions, and so forth. Despite these differences, the accuracy of error rate predictions was very similar for all projects.

Our results justify some optimism about the accuracy of PHRED quality scores for minor changes in sequencing technology, for example, sequences generated by new enzymes and fluorescent dyes. Initial studies showed that PHRED quality scores were also accurate for sequences produced by multiplex sequencing with radioactive detection (P. Richterich, unpubl.). However, we also observed two effects that can invalidate PHRED quality scores during these studies. First, sequences generated by chemical sequencing gave too low quality scores at mixed (A+G) reactions. Because secondary peak height is one of the parameters used in the error rate predictions, this is not surprising. Another potential source of error is high-frequency noise in the trace data. With such data, PHRED occasionally underestlinated the hand spacing by a factor of 2 or more, which resulted in incorrect base calls and quality scores. By applying simple smoothing algorithms to data with high-frequency noise, these problems could typically be resolved. Similar steps may be necessary to obtain accurate PHRED quality scores on data that have been generated by different sequencing instruments or preprocessed by different software.

Accurate quality scores can have a major impact on how sequences are used downstream from the sequence production process. In traditional sequencing projects where the goal is complete coverage at a final error rate below (e.g.) I in 10,000, the accuracy goals can be reached with single sequence reads as long as the quality scores are at least 40 (however, other potential problems like clone instability may make higher coverage advisable). Interesting questions arise as to how individual read quality contributes to project quality, or the error rate of the "final" sequence. Under the assumption that errors between different sequence reads are

completely independent, one could argue that two reads with a quality score of 20 (error probability of I in 100) are just as valuable as one sequence with a quality score of 40 (error probability of 1 in 10,000). However, although a single sequence stretch with quality levels above 40 would give a final sequence with an error rate of <1 in 10,000, assembling a consensus from two sequences with quality scores of 20 (1% error rate) could lead to one of two results: If the errors were completely random, the consensus sequence would be ambiguous at 2% of all locations; if the errors were completely localized, for example, because of reproducible compressions; the consensus sequence would have one "hidden" error every 100 bases. Typically, consensus sequences derived from low-quality sequences will have both kinds of problematic regions. Increased coverage can rapidly eliminate the random errors; however, increased coverage does not resolve errors from systematic sources. Manual examination of such problem areas is generally required; such "contig editing," however, tends to be time consuming, requires highly trained personnel, is an obstacle toward complete automation of DNA sequencing, and sometimes fails to climinate all errors. This leads to the somewhat counterintuitive conclusion that the practical value of increasing sequence qual-Ity can be even higher than indicated by the quality scores: One sequence of average quality above 40 can be "worth" more than two sequences of average quality 20.

Another application of DNA sequencing where high quality can be of disproportionately high value is the search for mutations in genomic DNA. In low quality sequences, secondary peaks and low resolution often complicate the identification of heterozygous mutations. In regions of higher sequence quality, such secondary peaks are smaller or absent and peaks are better resolved. Therefore, both falsepositive and false-negative errors can be significantly reduced in high-quality regions. Tools like PHRED, which can accurately measure sequence quality from trace data, can he of twofold value for mutation detection. First, base-specific quality scores can allow optimization of sequencing methods and strategies for mutation detection. Second, the quality scores can be used to evaluate the usefulness of individual sequence reads for mutation detection (e.g., by discarding reads below minimum thresholds); and they can guide software that automatically detects mutations.

The ability to predict error rates in a highly accurate fashion is likely to have a major impact in applications like those described above. PHRED is

the first widely used program that accurately predicts base-specific error probabilities. However, the algorithm for determining quality values has been described (Ewing and Green 1998), and it should be straightforward to implement similar quality values in other base-calling programs. Furthermore, an extension of the approach developed by Ewing and Green should be possible. For example, differentiation between mismatch and frameshift errors would enable better comparisons of sequencing methods with similar total error rates but different frameshift error rates. Several groups have described efforts to calculate separate probabilities (or "confidence assessments") for mismatch criors and frameshift errots (Lawrence and Solovyev 1994; Berno 1996). Their results demonstrated that different approaches to error type characterization are feasible and promising. Implementation of such error type predictions in other programs similar to the way PHRED uses quality scores would enable better method assessments, benchmarking, and production quality control, and could have a significant impact on downstream uses of DNA sequence information.

METHODS

Data Sets

For one project, sequence raw data in the form of ABI trace files were downloaded from a public FTP site. Sequence data for the five other projects were kindly provided by five different large-scale sequencing groups. Table I gives a summary of the six projects, and Table 2 gives an overview of the different sequencing methods used in the projects. The projects differed in the amount of prescreening of data that had been done, reflecting different approaches to quality control in different laboratories. In two projects (Il and C), different software programs had been used to identify and eliminate low-quality sequences. One project (F) included all data files generated, whereas the other three projects had excluded "falled lanes:"

Comparison of Actual and Predicted Error Rates

The sequences for all traces in each project were recalled using the program PHRED (v. 961028). Next, sequences in each project were assembled with PHRAP (P. Green, unpubl.). Slightly different methods were chosen for the statistical and graphical evaluation of the error rate prediction accuracy. In the statistical evaluation, only the longest contig produced by PHRAP was considered. The tables of aligned bases and observed discrepancy counts for

each quality score were taken from the PHRAP output and analyzed as follows. The expected number of discrepancies (E) at each quality score (q) was calculated by multiplying the number of aligned bases (N) with the error probability corresponding to the quality score: $E = N \cdot 10^{-0.10}$. The Spearman ranking coefficients were calculated by comparing the expected and observed error frequencies. To obtain the quantitative relation between the expected and observed error rates over the entire range, a least-

squares fit between the observed and expected rates was performed, with the intercept set to zero and the number of aligned bases at each quality score used as weights.

For a graphical comparison of estimated and actual error rates in 50-bp windows, the following steps were taken. For two of the projects, the consensus sequence was retrieved from public databases. For the four other projects, the DNA sequence and quality information were used by the program PHRAP to assemble consensus sequences for each of the projects. The individual reads were aligned to the consensus sequences of the longest contig, using the program CROSS_MATCH (P. Green, unpubl.), after removing single-coverage regions from the ends of the consensus sequence. CROSS-

MATCH uses an implementation of the Smith Waterman algorithm to generate alignments that typically do not include the ends of sequences, where disagreements are commonly due to vector

sequence or low quality sequence.

The quality files generated by PHRED and the alignment summaries generated by CROSS—MATCH were then analyzed as follows. First, the region of each query sequence that had been aligned by CROSS_MATCH was determined. Next, the actual and predicted error rates for the entire aligned part of each individual sequence was calculated. In addition, the average actual and predicted error rates for all alignable sequences together were calculated for windows of 50 bases in length. To calculate the predicted error rate, the quality scores g defermined by PHRED at each base were converted to error probabilities as described above (Ewing and Green 1998).

Subdividing Data into Subsets Based on Data Quality

To examine the accuracy of PHRED quality scores for data subsets of different quality within a project, the following approach was taken. For all sequence reads in project B, the number of bases with a quality score of at least 30 in each sequence was determined (bases with quality scores of at least 30 were called very high-quality bases, or VHQ bases). Se-

ESTIMATION OF ERRORS IN RAW DNA SEQUENCES

quences were sorted in descending order based on the number of very high-quality bases, and divided into four quartiles. Accordingly, quartile 1 contained 25% of sequences with the highest number of very high-quality bases, and quartile 4 contained the "worst" sequences. To illustrate the prediction accuracy in data with relatively high error rates, sequences from project B that had been "discarded" because they had not met the minimum quality criteria were added back to the data set. The sequences in each quartile were compared to the consensus sequences that had been generated using the entire data set, as described above for the graphical comparison.

Determining Actual Frameshift Error Rates

The calculation of actual frameshift error rates in the raw sequence data was performed using CROSS __MATCH, similar to the procedure described above for total error rates, except that only insertion and deletion errors were counted. Because PHRED does not give separate frameshift error estimates, a comparison of predicted and actual frameshift errors is not possible.

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REFERENCES

Berno, A.J. 1996. A graph theoretic approach to the analysis of DNA sequenting data. Genome Res. 6: 80-91.

Bonfield, J.K. and R. Staden. 1995. The application of numerical estimates of base calling accuracy to DNA sequencing projects. Nucleic Acids Res. 23: 1406-1410.

Churchill, G. and M.S. Waterman. 1992. The accuracy of DNA sequences: estimating sequence quality. *Genomics* 14: 89-98.

Ewing, B. and P. Green. 1998. Base-calling of automated sequencer traces using phred. II. Error probabilities. Genome Res. (this Issue).

Ewing, B., L. Hillier, M.C. Wendl, and P. Green. 1998. Base-calling of automated sequencer traces using placed. 1. Accuracy assessment. Genome Res. (this issue).

Lawrence, C.B. and V.V. Solovyev. 1994. Assignment of position-specific error probability to primary sequence data. Nucleic Acids Res. 22: 1272-1280.

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